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Moving from laboratory to real life conditions: Influence on the assessment of variability and stability of gait

This is the final peer-reviewed author's accepted manuscript (postprint) of the following publication:

*Published Version:*

Tamburini, P., Storm, F., Buckley, C., Bisi, M.C., Stagni, R., Mazzà, C. (2018). Moving from laboratory to real life conditions: Influence on the assessment of variability and stability of gait. *GAIT & POSTURE*, 59, 248-252 [10.1016/j.gaitpost.2017.10.024].

*Availability:*

This version is available at: <https://hdl.handle.net/11585/618931> since: 2019-01-28

*Published:*

DOI: <http://doi.org/10.1016/j.gaitpost.2017.10.024>

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(Article begins on next page)

This is the final peer-reviewed accepted manuscript of:

*Paola Tamburini, Fabio Storm, Chris Buckley, Maria Cristina Bisi, Rita Stagni, Claudia Mazzà, **Moving from laboratory to real life conditions: Influence on the assessment of variability and stability of gait**, Gait & Posture, Volume 59, 2018, Pages 248-252, ISSN 0966-6362*

The final published version is available online at:

<https://doi.org/10.1016/j.gaitpost.2017.10.024>

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**MOVING FROM LABORATORY TO REAL LIFE CONDITIONS:  
INFLUENCE ON THE ASSESSMENT OF VARIABILITY AND STABILITY  
OF GAIT**

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Submitted to

***Gait & Posture***

Word count:

2419

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**Acknowledgments**

ESB Mobility Award for young researcher (2016) and UK-EPSRC Frontier Engineering Awards (No.EP/K03877X/1) supported this study. The data used for this study are freely available through Figshare (DOI: 10.15131/shef.data.5519680).

## **Abstract**

The availability of wearable sensors allows shifting gait analysis from the traditional laboratory settings, to daily life conditions. However, limited knowledge is available about whether alterations associated to different testing environment (e.g. indoor or outdoor) and walking protocols (e.g. free or controlled), result from actual differences in the motor behaviour of the tested subjects or from the sensitivity to these changes of the indexes adopted for the assessment. In this context, it was hypothesized that testing environment and walking protocols would not modify motor control stability in the gait of young healthy adults, who have a mature and structured gait pattern, but rather the variability of their motor pattern.

To test this hypothesis, data from trunk and shank inertial sensors were collected from 19 young healthy participants during four walking tasks in different environments (indoor and outdoor) and in both controlled (i.e. following a predefined straight path) and free conditions. Results confirmed what hypothesized: variability indexes (Standard deviation, Coefficient of variation and Poincaré plots) were significantly influenced by both environment and walking condition. Stability indexes (Harmonic ratio, Short term Lyapunov exponents, Recurrence quantification analysis and Sample entropy), on the contrary, did not highlight any change in the motor control.

In conclusion, this study highlighted an influence of environment and testing condition on the assessment of specific characteristics of gait (i.e. variability and stability). In particular, for young healthy adults, both environment and testing condition affect gait variability indexes, whereas neither affect gait stability indexes.

**Key words:** daily life gait; variability indexes; stability indexes; indoor and outdoor walking; inertial sensors; accelerometers.

## Introduction

### TABLE NOMENCLATURE HERE

Laboratory assessment has been the standard setting for quantitative gait analysis for several decades. However, in recent years, the availability of wearable inertial measurement units has allowed to quantitatively and easily assess gait also out of the lab [1,2].

The assessment of gait out of the laboratory, whereby it is not constrained to a predefined path, aims at reproducing a testing condition more similar to that of daily living. This type of assessment is particularly interesting for the investigation of gait performance and of the underlying motor control with a specific focus on the quantification of dynamic stability and fall risk. It can potentially overcome the limitations (e.g. limited acquired number of stride) of data acquired in laboratory conditions [2]. Moreover, the monitoring of gait, as obtained from various types of quantitative descriptive indexes, provides information that can significantly impact the design of more effective training and rehabilitative interventions [3].

Several studies [2,4,5] analysed the gait pattern of faller and non-faller elderly and pathological subjects in daily-living conditions using indexes assumed to quantify the motor performance and the underlying motor control. However, limited knowledge is available regarding if and how the testing environment (e.g. indoor or outdoor) and the imposition of a specific walking path (e.g. free or controlled) might affect gait pattern and performance, and whether the indexes, commonly adopted to quantify these aspects, are sensitive to these changes.

Therefore, it is crucial to understand whether the alterations, associated to different testing conditions, result from actual differences in the motor behaviour of the analysed subjects or rather from the sensitivity of the indexes adopted for the assessment.

It is almost impossible to infer this knowledge analysing elderly and/or pathologic subjects, however for young healthy subjects, it can be assumed that the motor control of a mature and structured gait pattern [6], will not be significantly affected by the testing conditions. Therefore, environmental and testing conditions are not expected to modify the motor control stability in the gait of a young healthy adult, who has the

1 ability to face far more challenging conditions, but changes in the variability of the  
2 motor pattern could be expected as an adaptation to the environment **in order**  
3 to maintain stability.  
4

5 The definition and applicability of the concepts of variability and stability is well  
6 defined in mechanics, while the two are often used addressing similar meanings in  
7 **gait analysis referring to motor control. On one hand, in a complex dynamic system as**  
8 **human gait, variability could arise from the deterministic dynamics of the system (e.g.**  
9 **when a chaotic attractor is present as in human gait [7]).** It follows that the measured  
10 variability is a reflection of the multiple degrees of freedom of the system and does  
11 not necessarily imply destabilization of the system itself [7]. On the other hand,  
12 stability could arise from both the intrinsic properties of the system (i.e. motor  
13 control) and the specific movement pattern (i.e. gait) [7,8].  
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22 It could be argued that while gait variability is an indirect assessment of the  
23 motor control through gait performance (e.g. stride time), stability is instead a direct  
24 evaluation of the performance of the underlying motor control [7–9].  
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26

27 Besides traditional approaches based on the quantification of mechanical  
28 features of gait [10], a number of indexes have been proposed to quantify aspects  
29 more related to motor control [11,12]. These indexes can be generally grouped as  
30 variability (i.e. standard deviation, coefficient of variation, Poincarè plots) and  
31 stability indexes (i.e. Lyapunov exponents, harmonic ratio, sample entropy and  
32 recurrence quantification analysis) [11], based on their mathematical implementation  
33 and which characteristics of the analysed signal they are expected to quantify.  
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40 According to the above mentioned concepts of variability and stability,  
41 variability indexes, usually applied on stride time data, are meant to assess changes in  
42 the peripheral realization of the gait pattern [7,8,13,14], **whereas** stability indexes,  
43 usually applied on trunk acceleration data, are meant to assess the stability of the  
44 trajectory of the centre of mass. Indeed, recent studies [9,15], analysing both healthy  
45 (from 4 years-old children to 25 years-old young adults) and pathological subjects  
46 (stroke), analysed the role of the variability in joint kinematics in determining a  
47 **successful control of the stability of the centre of mass trajectory, approximated by**  
48 **the lower trunk [7,16,17]. Stride time and trunk acceleration data are two**  
49 **manifestations of the same control system in healthy and pathologic subjects [18,19].**  
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58 With this differentiation in mind, and since healthy young subjects have a well  
59 achieved and stabilized gait pattern [6], our hypothesis is that, when testing young  
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1 healthy subjects walking along both controlled and free paths, the indexes related to  
2 **motor stability are not expected to be significantly affected. Conversely,**  
3 modifications should be observed in variability indexes, due to the possibility to  
4 adjust the gait pattern to the environment in order to maintain stability.  
5

6  
7 In particular, an increase in variability indexes both from indoor to outdoor and from  
8 controlled to free conditions is expected, while no significant changes in gait stability  
9 indexes should be observed.  
10

11  
12 The present study aims at testing this hypothesis evaluating the influence of  
13 environment (indoor and outdoor) and testing conditions (controlled and free) on gait  
14 assessment when using variability and stability indexes in a young healthy population.  
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## 17 **Materials and Methods**

18  
19 **In a cross-over study,** nineteen healthy young volunteers (5 females, 14 males,  $28 \pm 3$   
20 years,  $1.75 \pm 0.09$  m,  $72.0 \pm 9.2$  kg) were recruited after having provided informed  
21 consent. Only subjects with no self-reported history of locomotor disturbances or  
22 injuries that could affect their normal walking behaviour, or cause fatigue during the  
23 experimental protocol were included in the study. The University of Sheffield's  
24 Research Ethics Committee granted ethical approval for the study.  
25

26  
27 Subjects wore two inertial measurement units (Opal, APDM, USA): one  
28 located on the lower trunk on the fifth lumbar vertebra, and one attached frontally on  
29 the right shank, 2 cm above the lateral malleolus, for stride detection [10]. Measures  
30 of accelerations of the trunk and angular velocity of the right shank were recorded at  
31 128 Hz.  
32

33  
34 Subjects completed four walking tasks in two different environments (indoor  
35 and outdoor) and in both controlled (i.e. following a predefined straight path) and free  
36 conditions (see details in Table 1) [20], indicated as ICW (Indoor Controlled  
37 Walking), OCW (Outdoor Controlled Walking), IFW (Indoor Free Walking) and  
38 OFW (Outdoor Free Walking), respectively. All participants performed the walking  
39 task in the different testing conditions, **in one day,** following the same order: OCW,  
40 OFW, IFW, ICW.  
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57 **TABLE 1 HERE (walking conditions)**  
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- ICW was performed in a quiet corridor of the university building, and participants were asked to walk on a straight line for 20 m. The distance was measured and marked on the floor using adhesive tape.
- For IFW condition, participants were instructed to walk inside the university corridors starting from the main entrance, with no restriction of route, opening and closing doors as necessary. The data was always collected during normal working hours, in mostly busy corridors.
- OCW was performed in a quiet open space within the university premises, on a flat tarmac surface.
- For OFW, participants were instructed to walk freely in the city centre, with no restrictions regarding route or walking speed, but avoiding stairs.

During IFW and OFW the participants did not have verbal interaction with other people, but they may have had to adjust their gait due to the presence of others in the surroundings. Interactions with other people were possible, particularly during IFW. Finally, during IFW and OFW turns could also be recorded in addition to straight walking. However, turns and resting periods were segmented and excluded from the analysis. Turn events with durations between 1-3 stride time and angles around the vertical axis over  $40^\circ$  were identified and removed using the method specified by El-Gohary et al. [21]. Resting periods were defined as those when the time between subsequent heel strikes [10] was higher than 1.5 s.

For each participant and each condition 80 strides were analysed, since this was the maximum number of strides available in all conditions.

Gait variability was assessed on stride times using the variability indexes:

- Standard Deviation (SD)
- Coefficient of Variation (CV) [22]
- Short term variability of stride estimated *via* Poincaré plots (PSD1) [23].

Gait stability was assessed applying to the vertical (v), medio-lateral (ml), and antero-posterior (ap) trunk acceleration components the stability indexes:

- Harmonic Ratio (HR), calculated decomposing the whole signal components into its harmonics (HR\_v, HR\_ap, HR\_ml) [24];
- Short term Lyapunov exponents (sLE) [25], calculated using the method defined by M.T. Rosenstein et al. [25]. The state space reconstruction was



1 composed by the delay embedded state spaces of each acceleration component  
2 (sLE\_v, sLE\_ml and sLE\_ap); data were not normalized.

- 3 • Recurrence quantification analysis (RQA) implying the calculation of  
4 recurrence rate (RR), determinism (DET) and averaged diagonal line length  
5 (AvgL) [26]  
6
- 7 • Sample entropy (SEN\_v, SEN\_ml and SEN\_ap), calculated for values of  $\tau$   
8 ranging from 1 to 6 according to methodology defined by previous analysis  
9 [27].  
10

11 These indexes were selected, among those previously used to detect changes in the  
12 gait pattern [4,12,28,29], based on the available number of consecutive strides per  
13 trial, which would ensure a reliability of at least 20% [11,30].

14 For calculation of sLE and RQA, the state space was constructed with an embedding  
15 dimension  $dE = 5$  and a time delay of 10 samples, as these parameters were defined  
16 appropriate for the analysis of gait data [31,32]. Raw unfiltered data were analysed to  
17 assure that information was not lost or altered.

18 Matlab R2015b (MathWorks BV, USA) was used for data and statistical  
19 analysis.

20 A Shapiro-Wilk test was performed on all the above-mentioned indexes,  
21 showing that they were not normally distributed. Median, 25<sup>th</sup> and 75<sup>th</sup> percentile  
22 values were hence calculated. Kruskal-Wallis test with minimum level of significance  
23 of 5% was performed to compare the indexes values obtained in the different walking  
24 conditions. Dunn-Sidak correction was considered for post-hoc analysis.  
25

## 26 **Results**

27 Figure 1 shows a representative time series of trunk acceleration in the antero-  
28 posterior direction and the angular velocity of the shank around the medio-lateral axis  
29 for each condition.  
30

## 31 **FIGURE 1 HERE**

32 All variability indexes varied significantly between the analysed walking  
33 conditions, conversely from the stability indexes (with the only exception of HR in  
34 both v and ap) as shown in Figure 2. In particular, the Kruskal-Wallis test showed  
35

1 statistically significant differences for PSD1 between OCW and OFW and between  
2 ICW and OFW, with values 35% higher in OFW than in OCW and ICW.

3 SD and CV in ICW were significantly different from both OCW and OFW  
4 conditions, being approximately 20% lower.  
5

6  
7 Despite the fact that HR\_v and HR\_ap significantly diminished when moving  
8 from ICW to OFW, the observed numerical differences were lower than the known  
9 reliability thresholds of this indexes [11]. Similarly, significant but not reliable  
10 variations were observed for HR\_ap between OFW and OCW and between OFW and  
11 IFW.  
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## 16 **FIGURE 2 HERE (BOX PLOT)**

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### 23 *Discussion*

24  
25 The walking pattern of young healthy adults was analysed in different environments  
26 (indoor and outdoor) and testing conditions (controlled and free) to assess if and how  
27 variability and stability, quantified using commonly used variability and stability  
28 indexes [11], would be affected. The hypothesis in the specific population was that  
29 stability would not change significantly, while variability would increase moving  
30 from indoor to outdoor and from controlled to free condition.  
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36 Overall, the results confirmed the study hypothesis: on one hand variability  
37 indexes, associated to the specific gait pattern, can be altered by testing conditions; on  
38 the other hand, stability indexes, related to the underlying motor control, are  
39 influenced neither by the environmental nor by the type of walking.  
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44 The differences observed in SD e CV values between Indoor Controlled  
45 Walking (ICW) and Outdoor Free Walking (OFW) indicate that stride time variability  
46 changes significantly when moving from the laboratory to outdoor walking  
47 conditions. This was further confirmed by the PSD1 values: both observed differences  
48 (ICW vs OFW and OCW vs OFW) and the trend (increased values from indoor  
49 controlled condition to the outdoor free one) highlighted how short-term variability of  
50 stride times [23] should be interpreted with caution when analysing data from  
51 different environments and testing conditions.  
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58 The variability indexes were influenced also by the environment in the controlled  
59 walking (ICW vs OCW): it has to be acknowledged that besides the change in the  
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1 environment, differences in the length of the path in the two walking conditions can  
2 also affect gait variability, as suggested for older subjects [33].

3  
4 No significant difference was found in stability indexes, in accordance with  
5 the study hypothesis. SEN showed similar values for all testing conditions, moreover  
6 the observed trends, in all directions, are in accordance to those reported in the  
7 literature [4,34]: higher SEN values for increasing  $\tau$ . This further supports the study  
8 hypothesis, highlighting how the stability of trunk acceleration during gait is not  
9 influenced by testing conditions and environment in young healthy adults.

10  
11 The increase in gait variability, associated to different testing conditions of  
12 gait in healthy young adults, suggests that this behaviour should not always be  
13 considered as a warning symptom, as usually interpreted for elderly subjects [35].  
14 Changes in variability are not necessarily related to a reduction of gait stability, hence  
15 not necessarily to be interpreted as an increase in fall risk.

16  
17 Given the many comparisons performed simultaneously, type I errors  
18 (multiple comparison problem) are deemed possible. However, in the present study,  
19 several comparisons were performed to investigate different aspects of gait control  
20 and, when assessing similar aspects (e.g. variability), the same trends were obtained  
21 from different parameters, thus reinforcing the results. In addition, a bias could have  
22 been introduced by the choice of performing the four walking tasks in the same order.  
23 However, both the homogeneity of the results (similar analysed aspects showed same  
24 behaviour) and, when present, highly significant differences ( $p\_value \ll 5\%$ ) suggest  
25 the potential bias to be marginal, if not negligible.

26  
27 In conclusion, this study highlighted the influence of environment and testing  
28 condition in the assessment of specific characteristics of gait (i.e. variability and  
29 stability). In particular, when assessing the gait of young healthy adults, both  
30 environment and testing condition affect variability indexes, whereas neither of the  
31 two affects stability indexes.

32  
33 In general, these results cannot be generalised to other populations, assuming  
34 that testing in or out of the lab will not affect gait stability assessment, for instance, in  
35 elderly and/or pathologic subjects. Nevertheless, the eventual assessment of  
36 significant differences in stability indexes, quantified for indoor and outdoor walking  
37 conditions in elderly and/or pathologic populations, would suggest an increased frailty  
38 of these subjects in terms of motor stability and fall risk, when compared to the  
39 reference performance of young healthy adults.

## Acknowledgements

ESB Mobility Award for young researcher (2016) and UK-EPSRC Frontier Engineering Awards (No.EP/K03877X/1) supported this study. The data used for this study are available through Figshare (10.15131/shef.data.5519680).

## References

- [1] B. J, U.W. Ebner-Priemer, J. Fahrenberg, Ambulatory activity monitoring: Progress in measurement of activity, posture, and specific motion patterns in daily life, *Eur. Psychol.* 14 (2009) 142–152. doi:10.1027/1016-9040.14.2.142.
- [2] A. Weiss, M. Brozgol, M. Dorfman, T. Herman, S. Shema, N. Giladi, J.M. Hausdorff, Does the evaluation of gait quality during daily life provide insight into fall risk? A novel approach using 3-day accelerometer recordings, *Neurorehabil. Neural Repair.* 27 (2013) 742–752. doi:10.1177/1545968313491004.
- [3] B.H. Dobkin, A. Dorsch, The Promise of mHealth, *Neurorehabil. Neural Repair.* 25 (2011) 788–798. doi:10.1177/1545968311425908.
- [4] E.A.F. Ihlen, A. Weiss, A. Bourke, J.L. Helbostad, J.M. Hausdorff, The complexity of daily life walking in older adult community-dwelling fallers and non-fallers, *J. Biomech.* 49 (2016) 1420–1428. doi:10.1016/j.jbiomech.2016.02.055.
- [5] S. Del Din, A. Godfrey, C. Mazzà, S. Lord, L. Rochester, Free-living monitoring of Parkinson’s disease: Lessons from the field, *Mov. Disord.* 31 (2016) 1293–1313. doi:10.1002/mds.26718.
- [6] Understanding motor development : infants, children, adolescents, adults / David L. Gallahue, John C. Ozmun. - Version details, Trove. (n.d.). <http://trove.nla.gov.au/version/36850867> (accessed November 30, 2016).
- [7] S.M. Bruijn, O.G. Meijer, P.J. Beek, J.H. van Dieën, Assessing the stability of human locomotion: a review of current measures, *J. R. Soc. Interface.* 10 (2013). doi:10.1098/rsif.2012.0999.
- [8] D. Hamacher, N.B. Singh, J.H. Van Dieën, M.O. Heller, W.R. Taylor, Kinematic measures for assessing gait stability in elderly individuals: a systematic review, *J. R. Soc. Interface.* 8 (2011) 1682–1698. doi:10.1098/rsif.2011.0416.
- [9] E. Papi, P.J. Rowe, V.M. Pomeroy, Analysis of gait within the uncontrolled manifold hypothesis: Stabilisation of the centre of mass during gait, *J. Biomech.* 48 (2015) 324–331. doi:10.1016/j.jbiomech.2014.11.024.
- [10] K. Aminian, B. Najafi, C. Büla, P.-F. Leyvraz, P. Robert, Spatio-temporal parameters of gait measured by an ambulatory system using miniature gyroscopes, *J. Biomech.* 35 (2002) 689–699. doi:10.1016/S0021-9290(02)00008-8.
- [11] F. Riva, M.C. Bisi, R. Stagni, Gait variability and stability measures: Minimum number of strides and within-session reliability, *Comput. Biol. Med.* 50 (2014) 9–13. doi:10.1016/j.combiomed.2014.04.001.

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- [12] D. Hamacher, D. Hamacher, N.B. Singh, W.R. Taylor, L. Schega, Towards the assessment of local dynamic stability of level-grounded walking in an older population, *Med. Eng. Phys.* 37 (2015) 1152–1155. doi:10.1016/j.medengphy.2015.09.007.
  - [13] E. Papi, P.J. Rowe, V.M. Pomeroy, Analysis of gait within the uncontrolled manifold hypothesis: Stabilisation of the centre of mass during gait, *J. Biomech.* 48 (2015) 324–331. doi:10.1016/j.jbiomech.2014.11.024.
  - [14] J.B. Dingwell, J. John, J.P. Cusumano, Do humans optimally exploit redundancy to control step variability in walking?, *PLoS Comput. Biol.* 6 (2010) e1000856. doi:10.1371/journal.pcbi.1000856.
  - [15] M.C. Bisi, R. Stagni, Is CoM kinematics a descriptive parameter of gait motor development? Verification on children and adults, *Gait Posture.* 42 (2015) S100. doi:10.1016/j.gaitpost.2015.06.182.
  - [16] W. Zijlstra, A.L. Hof, Displacement of the pelvis during human walking: experimental data and model predictions, *Gait Posture.* 6 (1997) 249–262. doi:10.1016/S0966-6362(97)00021-0.
  - [17] K. Wilmut, W. Du, A.L. Barnett, Gait patterns in children with Developmental Coordination Disorder, *Exp. Brain Res.* 234 (2016) 1747–1755. doi:10.1007/s00221-016-4592-x.
  - [18] W. Zijlstra, A.L. Hof, Assessment of spatio-temporal gait parameters from trunk accelerations during human walking, *Gait Posture.* 18 (2003) 1–10. doi:10.1016/S0966-6362(02)00190-X.
  - [19] J. McCamley, M. Donati, E. Grimpampi, C. Mazzà, An enhanced estimate of initial contact and final contact instants of time using lower trunk inertial sensor data, *Gait Posture.* 36 (2012) 316–318. doi:10.1016/j.gaitpost.2012.02.019.
  - [20] F.A. Storm, C.J. Buckley, C. Mazzà, Gait event detection in laboratory and real life settings: Accuracy of ankle and waist sensor based methods, *Gait Posture.* 50 (2016) 42–46. doi:10.1016/j.gaitpost.2016.08.012.
  - [21] M. El-Gohary, S. Pearson, J. McNames, M. Mancini, F. Horak, S. Mellone, L. Chiari, Continuous Monitoring of Turning in Patients with Movement Disability, *Sensors.* 14 (2013) 356–369. doi:10.3390/s140100356.
  - [22] J.M. Hausdorff, M.E. Nelson, D. Kaliton, J.E. Layne, M.J. Bernstein, A. Nuernberger, M.A.F. Singh, Etiology and modification of gait instability in older adults: a randomized controlled trial of exercise, *J. Appl. Physiol.* 90 (2001) 2117–2129.
  - [23] A.H. Khandoker, S.B. Taylor, C.K. Karmakar, R.K. Begg, M. Palaniswami, Investigating Scale Invariant Dynamics in Minimum Toe Clearance Variability of the Young and Elderly During Treadmill Walking, *IEEE Trans. Neural Syst. Rehabil. Eng.* 16 (2008) 380–389. doi:10.1109/TNSRE.2008.925071.
  - [24] H.B. Menz, S.R. Lord, R.C. Fitzpatrick, Acceleration patterns of the head and pelvis when walking on level and irregular surfaces, *Gait Posture.* 18 (2003) 35–46. doi:10.1016/S0966-6362(02)00159-5.
  - [25] M.T. Rosenstein, J.J. Collins, C.J. De Luca, A practical method for calculating largest Lyapunov exponents from small data sets, *Phys. Nonlinear Phenom.* 65 (1993) 117–134. doi:10.1016/0167-2789(93)90009-P.
  - [26] M.A. Riley, R. Balasubramaniam, M.T. Turvey, Recurrence quantification analysis of postural fluctuations, *Gait Posture.* 9 (1999) 65–78. doi:10.1016/S0966-6362(98)00044-7.

- 1 [27] M. Costa, C.-K. Peng, A. L. Goldberger, J.M. Hausdorff, Multiscale entropy  
2 analysis of human gait dynamics, *Phys. Stat. Mech. Its Appl.* 330 (2003) 53–60.  
3 doi:10.1016/j.physa.2003.08.022.
- 4 [28] M.C. Bisi, R. Stagni, Development of gait motor control: what happens after a  
5 sudden increase in height during adolescence?, *Biomed. Eng. OnLine.* 15  
6 (2016). doi:10.1186/s12938-016-0159-0.
- 7 [29] D. Hamacher, D. Hamacher, A. Törpel, M. Krowicki, F. Herold, L. Schega, The  
8 reliability of local dynamic stability in walking while texting and performing an  
9 arithmetical problem, *Gait Posture.* 44 (2016) 200–203.  
10 doi:10.1016/j.gaitpost.2015.12.021.
- 11 [30] B. Galna, S. Lord, L. Rochester, Is gait variability reliable in older adults and  
12 Parkinson’s disease? Towards an optimal testing protocol, *Gait Posture.* 37  
13 (2013) 580–585. doi:10.1016/j.gaitpost.2012.09.025.
- 14 [31] S.A. England, K.P. Granata, The influence of gait speed on local dynamic  
15 stability of walking, *Gait Posture.* 25 (2007) 172–178.  
16 doi:10.1016/j.gaitpost.2006.03.003.
- 17 [32] T.E. Lockhart, J. Liu, Differentiating fall-prone and healthy adults using local  
18 dynamic stability, *Ergonomics.* 51 (2008) 1860–1872.  
19 doi:10.1080/00140130802567079.
- 20 [33] B. Najafi, J.L. Helbostad, R. Moe-Nilssen, W. Zijlstra, K. Aminian, Does  
21 walking strategy in older people change as a function of walking distance?, *Gait*  
22 *Posture.* 29 (2009) 261–266. doi:10.1016/j.gaitpost.2008.09.002.
- 23 [34] M.C. Bisi, R. Stagni, Complexity of human gait pattern at different ages  
24 assessed using multiscale entropy: From development to decline, *Gait Posture.*  
25 47 (2016) 37–42. doi:10.1016/j.gaitpost.2016.04.001.
- 26 [35] J. Verghese, R. Holtzer, R.B. Lipton, C. Wang, Quantitative Gait Markers and  
27 Incident Fall Risk in Older Adults, *J. Gerontol. A. Biol. Sci. Med. Sci.* 64A  
28 (2009) 896–901. doi:10.1093/gerona/glp033.
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## *Abbreviations*

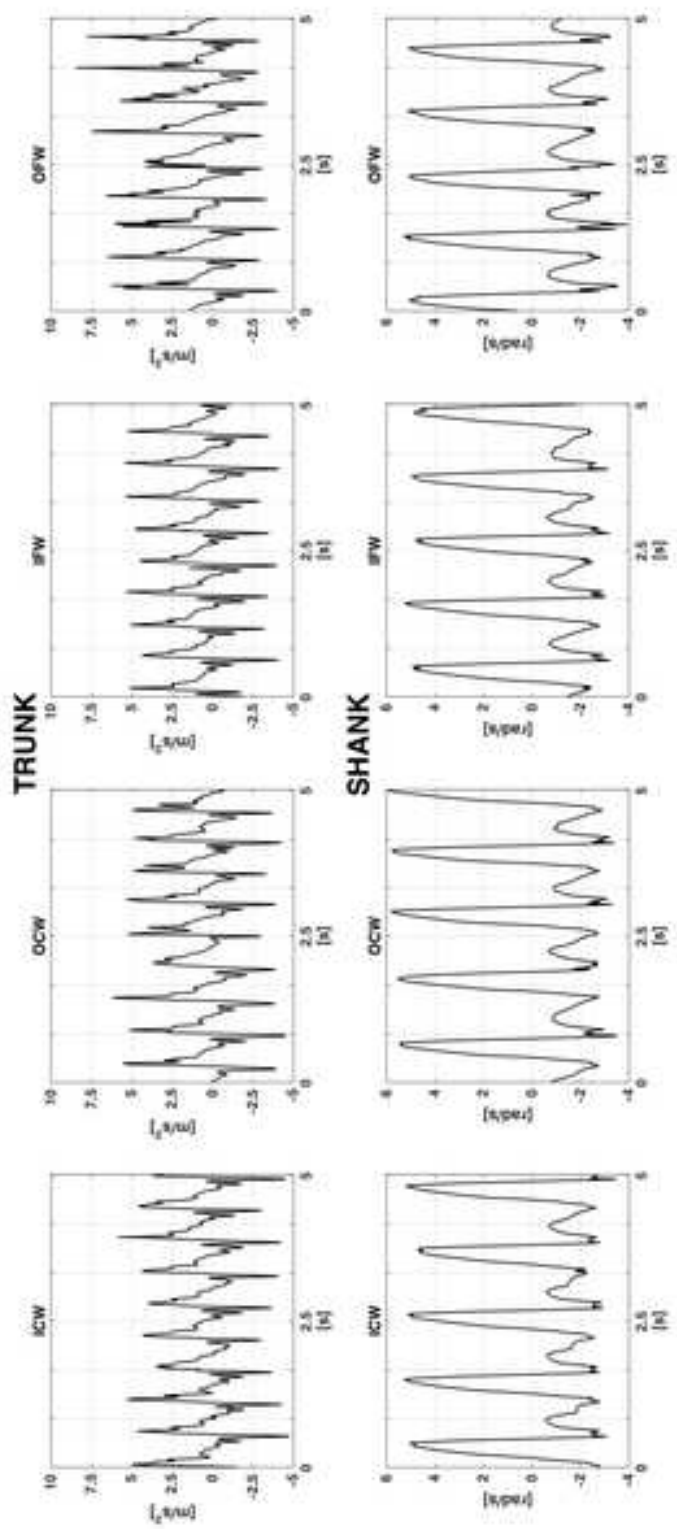
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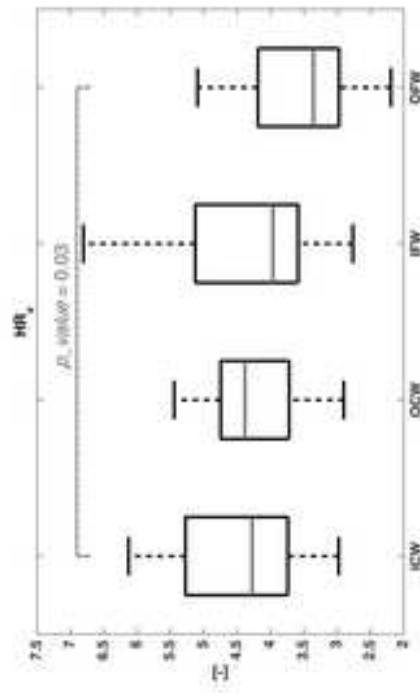
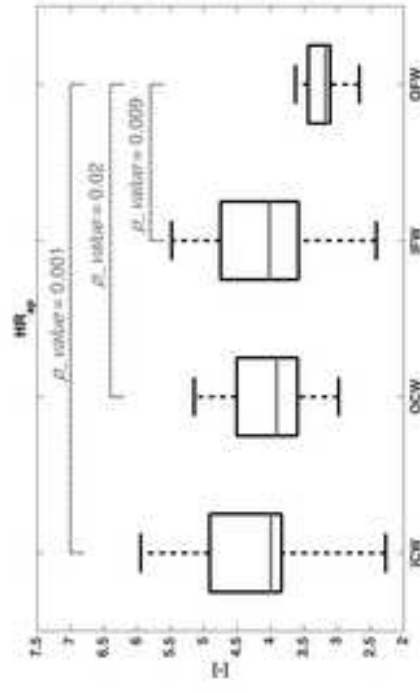
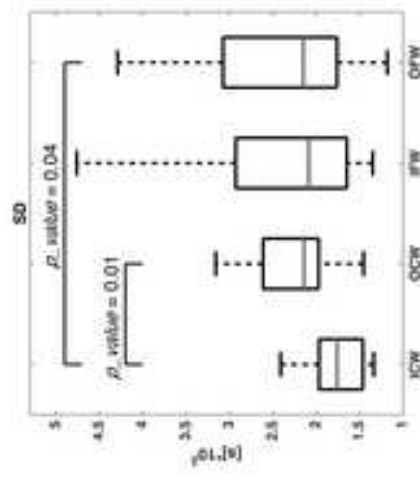
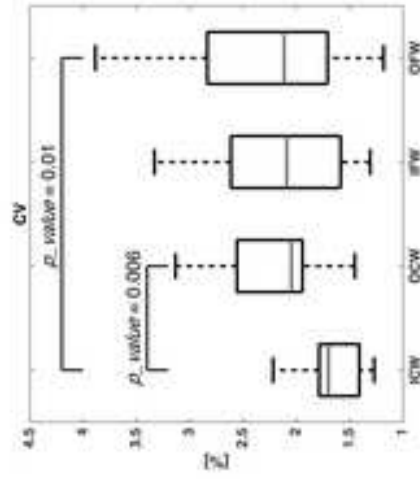
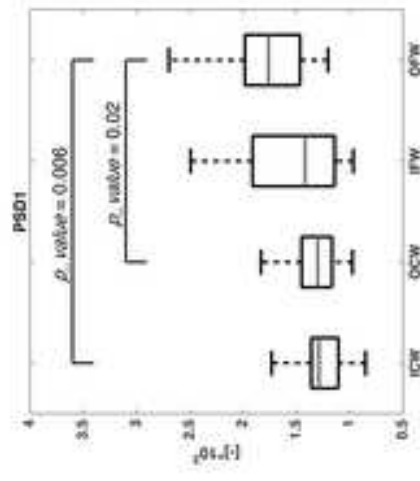
<b>SD</b>	Standard Deviation
<b>CV</b>	Coefficient of Variation
<b>PSD1</b>	Short term variability of stride estimated <i>via</i> Poincaré plots
<b>HR</b>	Harmonic Ratio
<b>HR_v</b>	Harmonic Ratio computed on the vertical acceleration direction
<b>HR_ml</b>	Harmonic Ratio computed on the medio-lateral acceleration direction
<b>HR_ap</b>	Harmonic Ratio computed on the antero-posterior acceleration direction
<b>sLE</b>	Short term Lyapounov Exponent
<b>sLE_v</b>	Short term Lyapounov Exponent on the vertical acceleration direction
<b>sLE_ml</b>	Short term Lyapounov Exponent on the medio-lateral acceleration direction
<b>sLE_ap</b>	Short term Lyapounov Exponent on the antero-posterior acceleration direction
<b>RQA</b>	Recurrence Quantification Analysis
<b>RR</b>	Recurrence Rate
<b>DET</b>	Determinism
<b>AvgL</b>	Averaged diagonal line length
<b>SEN</b>	Sample entropy
<b>SEN_v</b>	Sample entropy on the vertical acceleration direction
<b>SEN_ml</b>	Sample entropy on the medio-lateral acceleration direction
<b>SEN_ap</b>	Sample entropy on the antero-posterior acceleration direction

<b>Condition</b>	<b>Acronym</b>	<b>Description</b>	<b>Duration/Repetitions</b>
<b>Indoor controlled walking</b>	ICW	Walking at preferred speed along an indoor straight path 20.0 m long walkway.	Eight repetitions.
<b>Outdoor controlled walking</b>	OCW	Walking at preferred speed along an outdoor straight path 50.0 m long walkway.	Six repetitions.
<b>Indoor free walking</b>	IFW	Walking along corridors within a university building, avoiding stairs.	Two minutes.
<b>Outdoor free walking</b>	OFW	Walking along footpaths open to the public in the city centre without any restrictions in route or walking speed, avoiding stairs.	Fifteen minutes.

**Table 1:** Summary of the walking conditions performed during the experimental protocol, with acronym, description, and duration or repetition [20].







**Figure 1:** First line: representative trunk acceleration signal in antero-posterior direction in the four (ICW, OCW, IFW and OFW) walking conditions. Second line: representative angular velocity of the shank around the medio-lateral axis in the four (ICW, OCW, IFW and OFW) walking conditions.

**Figure 2:** Median, 25<sup>th</sup> and 75<sup>th</sup> percentiles and the *p\_value* of variability and stability indexes results showed significant differences (*p\_value*<5%) for the four walking conditions (ICW, OCW, IFW and OFW). Grey dotted lines indicate significant differences below the indexes reliability that can be associated to the analysed number of strides [11].

## Research highlights

Influence of testing conditions on gait variability and stability is unclear

Outdoor, indoor, controlled and free gait of healthy young adults were compared

Variability and stability indexes were applied on stride time and trunk acceleration

Variability indexes were influenced by both environment and type of walking

Stability indexes were influenced by neither environment nor type of walking