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Grimpampi, E., Oesen, S., Halper, B. et al. (3 more authors) (2015) Reliability of gait variability assessment in older individuals during a six-minute walk test. *Journal of Biomechanics*, 48 (15). pp. 4185-4189. ISSN 0021-9290

<https://doi.org/10.1016/j.jbiomech.2015.10.008>

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1 Reliability of gait variability assessment in older individuals during a 2 six-minute walk test

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16
17 **Abstract:** Gait variability is an important indicator of gait performance. However, the
18 reliability of the parameters used for its quantification, obtained from trunk linear accelerations,
19 has still not been thoroughly investigated. The aim of this study is to assess the reliability of gait
20 variability assessment in healthy older individuals based on lower trunk accelerations during a
21 six-minute walk test and to examine the reliability of the data acquired in shorter periods.
22 Twenty-nine subjects (84±5 years) performed the test while wearing one inertial sensor on the
23 lower trunk. Gait variability parameters (standard deviation and coefficient of variation of the
24 stride duration, and correlation coefficients of accelerations between neighbouring strides) were
25 calculated from the accelerations over 35 rectilinear strides observed during six series of one-
26 minute intervals extracted from the original signal. The reliability of these parameters was
27 assessed using intraclass correlation coefficients (ICC). Results showed no significant changes
28 across the six series for any of the parameters, with very high ICC values (0.93-0.95), indicating
29 a strong reliability of the observed quantities. Therefore, gait variability analysis based on lower

30 trunk acceleration data is a reliable and informative quantity in gait performance assessment in
31 older individuals, and one minute interval is sufficient to ensure reliable results.

32

33

34 **1. Introduction**

35 Gait in older population is characterised by progressive decrease in neuromotor control and
36 balance disorders (Hausdorff et al., 2001; Helbostad et al., 2007). Assessment of gait patterns
37 during normal overground walking is typically used for a better understanding of the postural
38 control system and of its responses in the presence of aging effects (Morris et al., 2001; Tyson,
39 1999).

40 Among the parameters used to investigate older people walking, gait variability has recently
41 gained popularity, being related to the underlying neural control of gait (Hausdorff, 2007). Its
42 analysis allows the identification of changes in the postural control system due to aging,
43 intervention or pathology, typically quantified through linear (standard deviation, SD, and
44 coefficient of variation, CV) (Lord et al., 2011) or non-linear (Buzzi et al., 2003) descriptors of
45 relevant variables.

46 The assessment of gait variability can be achieved using data measured with different
47 methods, e.g. instrumented walkways (Brach et al., 2010; Paterson et al., 2009), or inertial
48 measurement units (IMUs) (Annegarn et al., 2012; Riva et al., 2014). Using IMUs, it has been
49 shown that increased interstride variability in the sagittal plane and decreased variability of trunk
50 movements in the frontal plane, allow discriminating frailest older individuals (Moe-Nilssen and
51 Helbostad, 2005). Moreover, significantly increased variability along the medio-lateral direction
52 has been reported for patients with chronic obstructive pulmonary disease (Annegarn et al.,
53 2012). Gait pattern characteristics, as observed at trunk level, can hence be expected to be
54 clinically relevant.

55 Despite the widespread use of gait variability analysis, knowledge about its reliability is still
56 limited. This, together with the lack of standardized testing protocols, limits variability analysis
57 interpretation and understanding both for diagnostic and prognostic purposes in older population

58 (Lord et al., 2011). Reliable data can be obtained when investigating spatio-temporal parameters
59 variability by means of an instrumented walkway if analysing at least 35 steps (Galna et al.,
60 2013; Hollman et al., 2010). However, to the authors' knowledge, no specific indication is
61 available for the reliability of other gait parameters, such as those obtained from acceleration
62 signals, which allow to record the motion continuously over longer distances, thus potentially
63 resulting in more reliable estimates of gait variability. The purpose of this study is to assess the
64 reliability of gait variability parameters extracted from lower trunk acceleration signals during
65 the six-minute walk test (6MWT) performed by healthy elderly adults.

66

67 **2. Methods**

68 2.1. Participants

69 Twenty-nine elderly subjects, able to independently walk 10 metres, without neurological
70 disorders that could affect their performance and/or behaviour and with a Mini-mental state
71 examination score ($\geq 22/30$) (Folstein et al., 1975) participated in the study (Table 1), conducted
72 at three different retirement homes in Vienna. The chosen sample sized allowed to reach a
73 moderate effect size ($f=0.50$) with power=0.95 ($\alpha=0.05$) for a repeated measures ANOVA.
74 Informed consent was obtained from all participants.

75 [TABLE1]

76 2.2. Experimental procedure

77 A 6MWT was used to investigate gait variability. This test evaluates an individual's
78 functional capacity by measuring the distance a patient can walk in six minutes at their
79 maximum speed. Participants were asked to perform the 6MWT wearing their regular shoes
80 (Guyatt et al., 1985). They were asked to walk back and forth along a 30m straight pathway
81 (turning 180° at each end of the pathway) and to cover the maximum possible distance (6
82 minutes walk distance, 6MWD) by walking as fast as they could. An IMU (FreeSense, Sensorize

83 s.r.l Rome; fs=200 Hz) was positioned over their lower lumbar spine using an elastic belt
84 (Annegarn et al., 2012), which provided three linear acceleration and three angular velocity
85 components. This data allowed to perform both gait parameters and upper body acceleration
86 analyses.

87

88 2.3. Data analysis

89 Recorded signals were filtered with a 4th order Butterworth filter (cut-off frequency of 20 Hz)
90 (Mazzà et al., 2009). The IMU reference frame was rotated around the medio-lateral and antero-
91 posterior axes, as measured while the subject was standing upright, to align the unit local axes
92 with the three body anatomical axes (antero-posterior: AP, medio-lateral: ML, and vertical: V).
93 No further correction was applied to the data.

94 The acquired signals were initially segmented into 6 one-minute windows. Within each
95 window, the first rectilinear walking part was isolated using the gyroscope data measured around
96 the vertical axis to discard the turning parts. In each of these rectilinear parts, the peaks of the
97 antero-posterior accelerations (Zijlstra and Hof, 2003) were used to detect the beginning of a
98 stride cycle and to calculate the stride duration (T). The first series of 35 strides was found for
99 each of the 6 one-minute windows, and these six stride series were considered for further
100 analysis (Galna et al., 2013; Hollman et al., 2010).

101 The mean stride duration (M_T) and the root mean square values of each acceleration
102 component (RMS_V , RMS_{ML} , RMS_{AP}) were computed from the original signals and their
103 variation across the six stride series was observed to monitor changes in the overall walking
104 pattern that could be generated, for example, by fatigue.

105 The variability of the stride duration was calculated by combining interval information from
106 all strides within each series, and was assessed with the standard deviation (SD_T) and coefficient
107 of variation (CV_T). The mean values of the unbiased autocorrelation coefficients of the three
108 acceleration components, calculated over any two neighbouring strides (\overline{AC}_V , \overline{AC}_{ML} , \overline{AC}_{AP}),

109 were then computed to quantify between-stride acceleration variability using the method
110 proposed by (Moe-Nilssen and Helbostad, 2004). These coefficients were calculated from the
111 IMU data as follows:

112

$$113 \quad AC = \frac{1}{N-|m|} \sum_{i=1}^{N-|m|} x_i x_{i+m} \quad (1)$$

114

115 where: x_i are the samples of the acceleration signals ($i=1, \dots, N$) and m is the varying time lag
116 between the overlapped signal windows. The computation of the unbiased autocorrelation
117 coefficients was solved using the “xcov” algorithm (Matlab, MathWorks, Natick, MA) (Moe-
118 Nilssen and Helbostad, 2004). Perfect replication of the signals between neighbouring strides
119 would return $AC=1$. Large variations between neighbouring strides would give coefficients close
120 to 0.

121 2.4. Statistical analysis

122 Statistical analysis was implemented using SPSS software (version 20.0, SPSS Inc.). The
123 normal distribution of the analysed parameters (M_T , RMS_V , RMS_{ML} , RMS_{AP} , SD_T , CV_T , AC_V ,
124 AC_{ML} , AC_{AP}) was checked using Shapiro-Wilk’s test. Based on the normality test results,
125 parametric or non-parametric analysis was performed on gait variability parameters (SD_T , CV_T ,
126 AC_V , AC_{ML} , AC_{AP}) using a repeated measures ANOVA or a Friedman Test, respectively, for
127 each of the six stride series (within factor: 1st through 6th series, $\alpha=0.05$). The same analysis was
128 used for M_T , and RMS_V , RMS_{ML} , RMS_{AP} values to check for variations in the overall gait
129 pattern (e.g. due to possible fatigue). Post-hoc t-test analysis with Bonferroni correction was
130 used when significant differences were found. Intraclass correlation coefficients (ICC, two
131 factor, mixed effects model) were used to assess within-session reliability for each gait
132 variability parameter measured over all six stride series.

133 The correlation between gait variability measures and subjects' functional capacity was
134 assessed with the Pearson's (R) correlation coefficient between the computed variability
135 parameters and the 6MWD.

136

137 **3. Results**

138 All the subjects were able to complete the 6MWT, with an average 6MWD of 330 ± 75 m,
139 which is much lower value than that typical of healthy individuals (Du et al., 2009; Enright and
140 Sherrill, 1998). No statistically significant differences between the six stride series were found
141 for the gait variability parameters (SD_T , CV_T , AC_V , AC_{ML} , AC_{AP}). For the M_T , a significant
142 difference was found among the six stride series ($F_{5,28}=18.437$, $P<0.001$). Post-hoc testing
143 revealed that M_T increased by 5.5% between the 1st and the 6th series (t-test, $p<0.001$). A
144 significant decrease was observed in RMS_V ($\chi^2_{28}=31.327$, $P<0.001$) and RMS_{AP} ($\chi^2_{28}=13.847$,
145 $P<0.05$) during the 6MWT (Table 2).

146

[TABLE2]

147 Descriptive statistics of the temporal and trunk acceleration gait variability parameters is
148 presented in Table 3 (mean and SD) and Figure 1 (median and quartiles of temporal parameters)
149 for each of the six series of strides. No significant changes and strong reliability values were
150 found during the 6MWT (ICC equal to 0.95 and 0.94 for SD_T and CV_T , respectively).

151

[TABLE3]

152

[FIGURE1]

153 Figure 2 shows the descriptive statistics of the parameters extracted from the trunk
154 acceleration signals for the six stride series. No significant changes and a strong reliability was
155 found for all the acceleration components, with ICCs values equal to 0.93, 0.95 and 0.93 for
156 \overline{AC}_V , \overline{AC}_{ML} , and \overline{AC}_{AP} respectively. Moreover, across all stride series \overline{AC}_{AP} and \overline{AC}_V were
157 significantly higher than \overline{AC}_{ML} (t-test, $p<0.001$).

158 [FIGURE2]

159 The 6MWD was found to present a significant negative correlation with SD_T ($R=-0.57$,
160 $p=0.001$) and CV_T ($R=-0.60$, $p=0.001$) and a significant positive correlation with \overline{AC}_V ($R=0.57$,
161 $p=0.001$).

162

163 **4. Discussion**

164 The aim of this study was to assess the reliability of gait variability parameters as extracted
165 from lower trunk acceleration signals in a group of healthy elderly subjects during a 6MWT.
166 Reported results showed no significant changes during the six minutes for any of the
167 investigated gait variability parameters, with reported ICC values indicating a strong reliability
168 (Shrout and Fleiss, 1979) of all the observed quantities. The generalisation of these results to
169 data obtained from sensors located on different parts of the body would require further
170 investigations.

171 Reported variability values were similar to those previously reported for geriatric subjects
172 (van Iersel et al., 2007) and remained reliable throughout the six minutes. This was true despite
173 the fact that the 6MWT might have actually fatigued the subjects (Kervio et al., 2003), as
174 suggested by the increase in the average stride duration observed in the last minute of the trial.
175 Further study would be needed to investigate more thoroughly this assumption and prove the
176 robustness of gait variability to fatigue.

177 Results obtained by calculating stride variability from 35 strides confirmed those obtained with
178 an instrumented walkway (Galna et al., 2013), suggesting that highly reliable assessment of
179 stride-to-stride fluctuations do not require prolonged acquisitions. The same study (Galna et al.,
180 2013) suggested that two minutes are needed to reliably assess gait variability through spatio-
181 temporal parameters. Our results indicate that, when variability analysis is based on lower trunk
182 acceleration data, 35 strides performed in a shorter period, less than one minute in our case, are

183 sufficient to assess gait variability with the same reliability as the 6MWT. This is especially
184 important for older persons as it implies that testing could be performed with less physical
185 demand.

186 Trunk acceleration variability values were also in accordance with those previously reported
187 for similar subjects during tests of shorter duration (Annegarn et al., 2012; Moe-Nilssen and
188 Helbostad, 2005). In particular, it was confirmed that across all stride series \overline{AC}_{AP} and \overline{AC}_V were
189 significantly higher than \overline{AC}_{ML} (Helbostad et al., 2007), indicating that larger variations were
190 found between neighbouring strides in the ML direction. It has been suggested (Annegarn et al.,
191 2012; Mazzà et al., 2008; Moe-Nilssen and Helbostad, 2005) that variability in the ML direction
192 may be related to different balance control mechanism than variability in the two other
193 directions. Further studies are needed to elucidate whether this is solely the result of a lower
194 signal-to-noise ratio associated with the acceleration signals in the ML components typically
195 being the lowest in amplitude.

196 A limitation of this study is that, due to the limited number of collected consecutive strides, the
197 assessment of variability was only based on linear techniques. Further studies, including the
198 collection of longer stride series, would be required to include non-linear analysis tools. A
199 further limitation might be that the subjects knew beforehand that the task duration was of six
200 minutes and might have adjusted their walking strategy accordingly. Despite this hypothesis
201 might be discarded according to indications available in the literature (Kosak and Smith, 2005),
202 further studies might be needed to elucidate this aspect.

203 The reported results suggest that gait variability is a suitable assessment of elderly subjects'
204 gait performance. Both temporal and trunk acceleration parameters, in fact, were correlated to
205 the functional capacity of the subjects with the subjects who were able to walk further being
206 those with smaller stride time variability and smaller vertical trunk acceleration variability.
207 However, these correlations were only moderate and further studies on a larger sample are
208 needed to draw stronger conclusions.

209 According to the reported results, it can be concluded that: a) gait variability, as measured
210 both in terms of temporal and trunk acceleration parameters by an IMU located on the lower
211 trunk, is a reliable and informative quantity in the assessment of gait performance in healthy
212 elderly subjects; b) a shorter version of the 6MWT, reduced to one minute, allows to reliably
213 assess gait variability, ensuring less physical demand on the elderly population.

214

215 **Conflict of interest statement**

216 The authors have no conflicts of interest to report.

217

218 **Acknowledgments**

219 This work has been supported by funds from the Italian Ministry of Education and Research
220 (PRIN 2010R277FT, “*Fall risk estimation and prevention in the elderly using a quantitative*
221 *multifactorial approach*”). The support of the Research Platform Active Ageing (University of
222 Vienna), and of the University of Rome “Foro Italico” is also gratefully acknowledged.

223 Ethical Approval: The present study was conducted in accordance to the Austrian laws
224 (including doctors Act, CISA, Data Protection Act), the Declaration of Helsinki (as revised in
225 Edinburgh 2000) and in analogous accordance with ICH-GCP Guidelines. Written informed
226 consent was obtained from all participants. This study was approved by the ethics committee of
227 the City of Vienna (EK-11-151-0811) and registered at ClinicalTrials.gov, NCT01775111.

228

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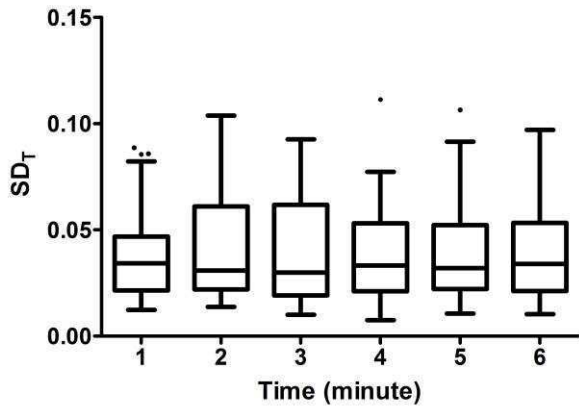
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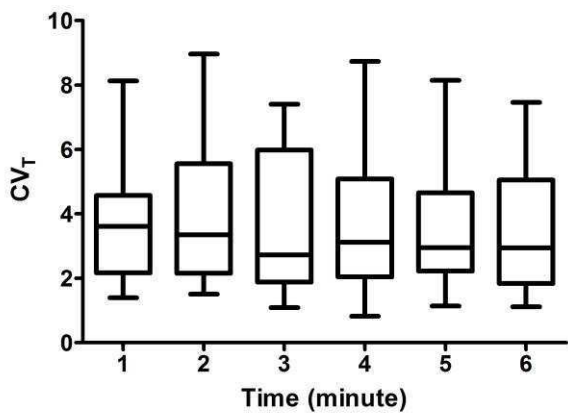
295 **Figures**

296

297 **Fig. 1.** Box-plots showing minimum, lower quartile, median, upper quartile, maximum, and
298 outliers of: (a) standard deviation (SD_T), and (b) coefficient of variation (CV_T) of stride
299 durations during the six stride series.



(a)

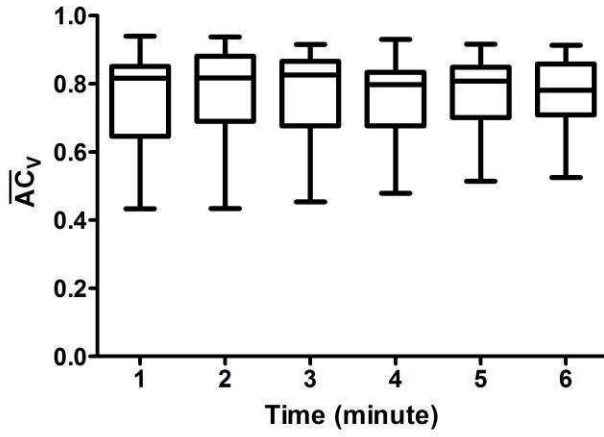


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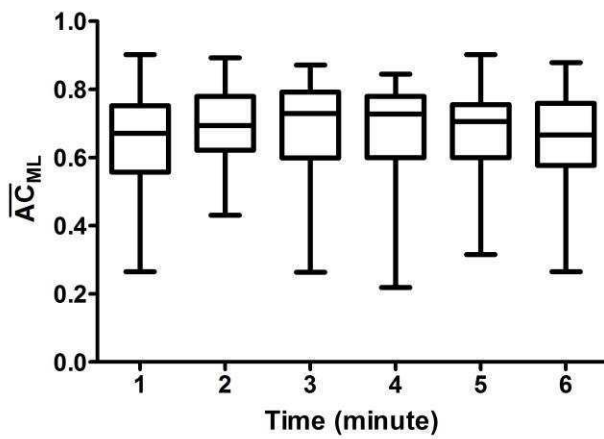
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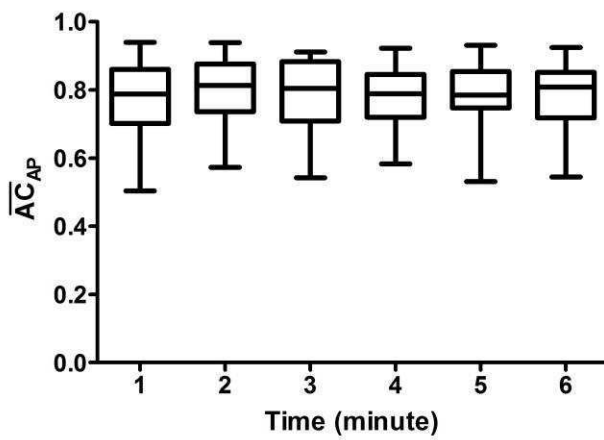
302 **Fig. 2.** Box-plots showing minimum, lower quartile, median, upper quartile, maximum, and
 303 outliers of interstride trunk variability AC along: (a) vertical: \overline{AC}_V , (b) medio-lateral: \overline{AC}_{ML} , and
 304 (c) antero-posterior: \overline{AC}_{AP} anatomical body axes during the six stride series.



(a)



(b)



(c)

305

306

307 **TABLE 1** Anthropometric characteristics of subjects. Value expressed as mean \pm standard
 308 deviation

N = 29	
Women	24
Men	5
Age (years)	84 \pm 5
Mass (kg)	75 \pm 15
Height (m)	1.59 \pm 0.08
BMI (kg/m ²)	30 \pm 6

309
 310 **TABLE 2** Mean stride duration (M_T) and acceleration root mean square (RMS) of each axis (V:
 311 vertical, ML: medio-lateral, AP: antero-posterior) during the six-minute walk test (6MWT). Data
 312 are expressed as median \pm IQR except for M_T reported as mean \pm standard deviation (n = 29).
 313

	1 st series	2 nd series	3 rd series	4 th series	5 th series	6 th series	P
M_T	1.03 \pm 0.12	1.06 \pm 0.11	1.07 \pm 0.11	1.08 \pm 0.12	1.09 \pm 0.12	1.09 \pm 0.13	<0.001
RMS _V	1.95 \pm 0.96	1.82 \pm 0.93	1.88 \pm 1.04	1.87 \pm 0.88	1.83 \pm 0.85	1.81 \pm 0.93	<0.001
RMS _{ML}	1.31 \pm 0.62	1.28 \pm 0.58	1.28 \pm 0.58	1.32 \pm 0.60	1.30 \pm 0.67	1.33 \pm 0.58	0.759
RMS _{AP}	1.23 \pm 0.49	1.19 \pm 0.44	1.18 \pm 0.42	1.17 \pm 0.41	1.15 \pm 0.42	1.16 \pm 0.37	0.017

314
 315 **TABLE 3** Standard deviation (SD_T) and coefficient of variation (CV_T) of the stride duration and
 316 mean values of interstride trunk variability (\overline{AC}) along the three anatomical body axes, for each
 317 of the six series of strides (n = 29). Value expressed as mean \pm standard deviation.

(s)	1st series	2nd series	3rd series	4th series	5th series	6th series
SD_T	0.04 \pm 0.02	0.05 \pm 0.03	0.04 \pm 0.02	0.04 \pm 0.02	0.04 \pm 0.02	0.04 \pm 0.02
CV_T	3.85 \pm 2.19	4.18 \pm 3.02	3.83 \pm 2.20	3.59 \pm 1.95	3.61 \pm 1.98	3.48 \pm 1.78
\overline{AC}_V	0.74 \pm 0.15	0.78 \pm 0.12	0.77 \pm 0.12	0.76 \pm 0.11	0.78 \pm 0.11	0.77 \pm 0.10
\overline{AC}_{ML}	0.65 \pm 0.15	0.69 \pm 0.12	0.68 \pm 0.16	0.67 \pm 0.16	0.68 \pm 0.14	0.66 \pm 0.14
\overline{AC}_{AP}	0.77 \pm 0.10	0.79 \pm 0.09	0.78 \pm 0.10	0.78 \pm 0.10	0.78 \pm 0.10	0.78 \pm 0.10

319