

Surface effects on dynamic stability and loading during outdoor running using wireless trunk accelerometry

Kurt H. Schütte ^{a, c*}, Jeroen Aeles ^a, Tim Op De Beéck ^b, Babette C. Van der Zwaard ^c,

Rachel Venter ^c, Benedicte Vanwanseele ^a

^aHuman Movement Biomechanics Research Group, Department of Kinesiology, KU Leuven, Leuven, Belgium.

^bDepartment of Computer Science, KU Leuven, Leuven, Belgium.

^cMovement Laboratory, Department of Sport Science, Stellenbosch University, Stellenbosch, Western Cape, South Africa.

***Corresponding author:** Email: kurtschutte.sa@gmail.com

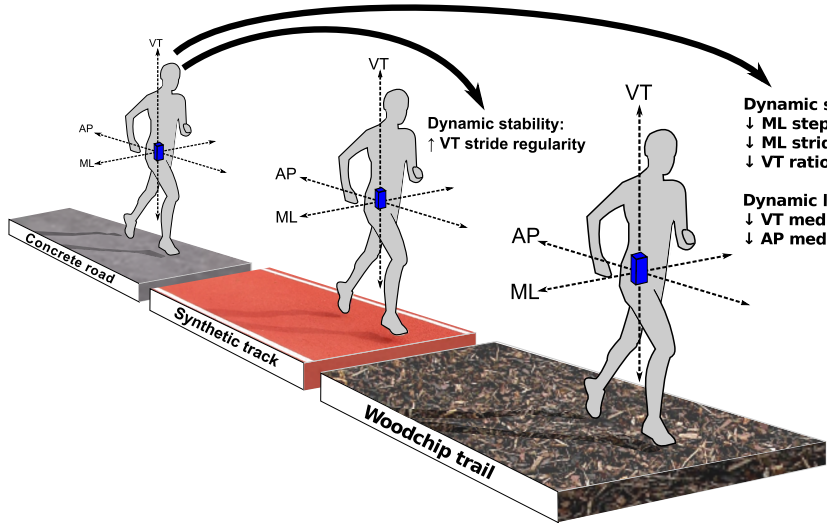
Word count

Abstract: 220

Body of text: 3193

Acknowledgments

The authors would like to acknowledge the contributions of Hannelore Boey, Sophie Plessers, and Tine Ravyts to data collection of this research. Funding: K.H.S received doctoral scholarships from the National Research Foundation (ZA) (# SFH13071922430, <http://www.nrf.ac.za/>) and the Erasmus Mundus scholarship of the European Union (<http://www.ema2sa.eu/>).



Dynamic stability:
↑ VT stride regularity

Dynamic stability:
↓ ML step regularity (symmetry)
↓ ML stride regularity
↓ VT ratio of acceleration RMS

Dynamic loading:
↓ VT median frequency of impact
↓ AP median frequency of impact

ABSTRACT

Despite frequently declared benefits of using wireless accelerometers to assess running gait in real-world settings, available research is limited. The purpose of this study was to investigate outdoor surface effects on dynamic stability and dynamic loading during running using tri-axial trunk accelerometry. Twenty eight runners (11 highly-trained, 17 recreational) performed outdoor running on three outdoor training surfaces (concrete road, synthetic track and woodchip trail) at self-selected comfortable running speeds. Dynamic postural stability (tri-axial acceleration root mean square (RMS) ratio, step and stride regularity, sample entropy), dynamic loading (impact and breaking peak amplitudes and median frequencies), as well as spatio-temporal running gait measures (step frequency, stance time) were derived from trunk accelerations sampled at 1024 Hz. Results from generalized estimating equations (GEE) analysis showed that compared to concrete road, woodchip trail had several significant effects on dynamic stability (higher AP ratio of acceleration RMS, lower ML inter-step and inter-stride regularity), on dynamic loading (downward shift in vertical and AP median frequency), and reduced step frequency ($p < 0.05$). Surface effects were unaffected when both running level and running speed were added as potential confounders. Results suggest that woodchip trails disrupt aspects of dynamic stability and loading that are detectable using a single trunk accelerometer. These results provide further insight into how runners adapt their locomotor biomechanics on outdoor surfaces *in situ*.

Keywords: Running gait, running surface; trunk accelerometer; dynamic stability; dynamic loading.

1 **Introduction**

2 Worldwide millions of people participate in recreational and competitive running. It is an easily
3 accessible sport with numerous proven health benefits. However, repetitive collisions with the
4 ground also make running a sport with a high incidence of chronic overload injuries [1].

5 Dynamic loading related variables such as higher vertical loading rates [2] or peak tibial
6 accelerations [3] have been prospectively associated with lower-limb overuse running injuries
7 such as stress fractures. It is commonly believed that these dynamic loads and subsequently
8 overuse injury risk is exacerbated on harder surfaces such as concrete or asphalt. However,
9 epidemiological research has thus far failed to find any relationship between surface hardness
10 and injury, possibly due to difficulty in accurately quantifying time and intensity on typical
11 running surfaces [4]. Identifying how dynamic loads are moderated on typical running surfaces
12 could therefore add insights into appropriate preventative strategies for overuse running injury.

13 Laboratory studies have shown that small alterations in running surface can induce changes in
14 human running mechanics. For example, it is known that softer [5–7] or uneven [8,9] running
15 surfaces cause runners' to rapidly increase their leg stiffness, while peak ground reaction forces
16 are mostly moderated with a stable centre of mass (CoM) trajectory [5–7]. Although, Dixon et al.,
17 [10] reported individual specific adaptations in knee kinematics between asphalt and a softer
18 rubber-modified surface, they [10] also observed an overall reduction in vertical loading rates
19 when switching to the softer surface. While these aforementioned studies provide essential
20 insights, the mechanisms for moderating are perhaps not directly applicable to “real-world”
21 running surfaces that naturally vary in composites of hardness, evenness, and gradient.

22 In attempt to secure ecological validity, some researchers have investigated how runners adapt
23 their loading and mechanics to typical outdoor running surfaces. Using cinematography, Creagh
24 et al., [11] found that running in long grass decreased step lengths while increased hip vertical
25 displacement, knee lift and peak upper leg angles compared to running on tarmac. Others who
26 have used portable wearable devices such as in-shoe plantar pressure measurements or tibial

1 accelerometry have found conflicting results. For example, Tessutti et al., [12] reported higher
2 central and lateral peak plantar pressures along with shorter contact times when running on
3 asphalt compared to natural grass. In contrast, no differences in peak plantar pressure [13],
4 impulse [14], tibial shock [13] or contact times [13,14] have been found between concrete,
5 grass, or synthetic track. Discrepancies in findings could be attributed to large inter-individual
6 responses [10]. It appears that there is a need for a better understanding of how runners
7 moderate their loading and gait in response to “real-world” surfaces.

8 Measures derived from wireless tri-axial trunk accelerometers have become a popular approach
9 to reliably and unobtrusively capture dynamic loading and CoM stability in various
10 environments. The acceleration root mean square (RMS) as well as the autocorrelation-based
11 coefficients referred to as inter-step and inter-stride regularity have identified a wide variety of
12 impaired or asymmetrical stability patterns related to ageing [15], lower limb prosthesis [16],
13 hemiplegia [17], and gross motor function [18]. When applied to running gait, these measures
14 can detect compensations in dynamic stability due to fatigue [19,20], predict oxygen
15 consumption [21], and classify athletes based on their training background [22]. The current
16 paper includes stability and impact frequency components of running gait, which may be more
17 sensitive to changes in surface relative to other measures i.e. spatio-temporal or impact peaks.
18 However, these accelerometer measures have usually been investigated on a single running
19 surface, thus limiting multi-terrain generalizability.

20 Woodchip trails are becoming popular running surfaces that are specifically constructed to have
21 “structural dampening” to reduce impact-loading related injuries and enhance participation of
22 recreational running. Indeed, animal studies suggest that woodchip surfaces have injury
23 preventative properties. For example, adult sheep that were exposed to prolonged activities on
24 woodchips were less prone to development of knee osteoarthritis compared to sheep exposed to
25 activities on hard concrete [23]. In addition, hoof impact accelerations were significantly more
26 dampened when horses trotted at $\sim 4 \text{ m}\cdot\text{s}^{-1}$ on woodchip surface compared to asphalt [24].

1 Unfortunately, previous research on human running has primarily focused on other outdoor
2 surfaces such as grass [11–14], with no apparent evidence on woodchip trails. The purpose of
3 this study was to investigate outdoor surface effects on dynamic stability and loading during
4 running using tri-axial trunk accelerometry. Based on previous laboratory research indicating
5 smoothness of CoM trajectory under different surface conditions, we hypothesized that trunk
6 accelerometry measures of dynamic stability and loading would be minimally affected by
7 running surface.

8 **Methods**

9 *Participants*

10 Two predetermined age-matched groups of endurance runners aged 18 to 33 years of mixed
11 gender (# women 14, 50 %) were recruited for this study; highly-trained runners (mileage > 50
12 km·week⁻¹, n = 13) and recreational runners (mileage < 30 km·week⁻¹, n = 17). All participants
13 were screened to have no history of lower extremity injury within the past three months.
14 Written informed consent was received from all runners prior to participation in accordance
15 with the Declaration of Helsinki. The study was approved by the local ethics committee
16 (Commissie Medische Ethiek KU Leuven).

17 *Experimental protocol*

18 All runners (n = 17 recreational; n = 13 highly-trained) performed a standardized warm-up.
19 Outdoor running was performed on 90m of straight and flat concrete road, synthetic track, and
20 woodchip trail. Photo electronic timing gates (RaceTime 2 system, Microgate, Bolzano, Italy)
21 were positioned to capture average running speed from the 10m to 70m mark. A practice trial
22 was provided to familiarize participants to each surface. The self-selected running speed on
23 concrete was used as control speed on the other surfaces, and trials on subsequent running
24 surfaces were discarded if the running speed differed by $\pm 1 \text{ m}\cdot\text{s}^{-1}$ of control speed. The order of
25 the other two surfaces was randomized. To avoid any fatigue effect runners were allowed to rest
26 during five minutes between each surface.

1 *Accelerometry measurements*

2 Tri-axial accelerometry (X50-2 wireless accelerometer, range $\pm 50g$, sampling at 1024 Hz,
3 0.016g/count resolution, 33g weight, Gulf Coast Data Concepts, MS, USA) was acquired during
4 each running trial. The accelerometer was securely positioned over L3 spinous process of the
5 trunk [25], and directly mounted to the skin using double sided tape and adhesive spray.
6 Accelerometer position was unaltered between all running trials and was routinely checked
7 between running trials for security. Trials were discarded in the case the investigators deemed
8 the accelerometer to be not securely fastened upon its removal (after data collection).

9 All signal processing of acceleration curves was performed using customized software in
10 MATLAB version 8.3 (The Mathworks Inc., Natick, MA, USA). Accelerometry-derived parameters
11 were computed from the middle ten consecutive strides of the 10-70m measurement zone, that
12 were first trigonometrically tilt-corrected and filtered using a zero-lag 4th order low-pass
13 Butterworth filter (cut-off frequency 50 Hz) [20,25]. Accelerometry-derived parameters were
14 averaged over two running trials per surface per participant.

15 *Outcome measures*

16 Spatio-temporal parameters were quantified by step frequency and stance time. The former was
17 acquired using the time lag of the first dominant peak of the vertical acceleration's unbiased
18 autocorrelation [20,25]. The latter was acquired based on the heuristic that as long as the body
19 is accelerating upwards, the foot should still be in contact with the ground [26]. Therefore, zero
20 crossings of vertical accelerations identified periods where the vertical acceleration was
21 positive and accelerating upwards (initial contact to final contact) [26].

22 Dynamic postural stability parameters were quantified from tri-axial (vertical, ML, AP)
23 accelerations firstly using the ratio of each linear acceleration axis root mean square (RMS)
24 relative to the resultant vector RMS to capture variability [21]; secondly using step and stride
25 regularity (unbiased autocorrelations procedure) to capture symmetry and consistency of
26 running steps and strides respectively, with perfect regularity equivalent to one [25]; and thirdly

1 using sample entropy from raw accelerations to capture the waveform predictability, with
2 higher values indicating less periodicity or more unpredictability [27]. Detailed procedures and
3 algorithm inputs for the computation and extraction of these dynamic postural stability
4 parameters are explained previously [20].

5 Dynamic loading parameters during stance were computed from extracted stance phases firstly
6 in the time domain and secondly in the frequency domain. The former was acquired by
7 extracting the peak positive vertical (impact) and peak negative anteroposterior (breaking)
8 accelerations identified between 1% and 20% stance. The latter was acquired from the median
9 frequency of vertical and AP accelerations of the entire stance phase calculated as the centroid of
10 the power spectral density (PSD) curves within the 1 – 100 Hz range [28]. PSD was calculated
11 from the Fast Fourier Transform (FFT) of unfiltered vertical stance phase accelerations from 0
12 to the Nyquist (F_N) frequency, that were first processed in line with previous methods [29]:
13 detrended, padded with zeros to equal 2048 data points (ensuring 2^n periodicity), and
14 interpolated to 1 Hz bins.

15 *Statistical analysis*

16 Group descriptive characteristics were compared using independent t-tests. Each
17 accelerometry-derived parameter was individually evaluated for normality; skewness between
18 >-1 and <1 was accepted. Subsequently, normally distributed data was analyzed by means of
19 linear regression using generalized estimating equations (GEE). GEE analysis is more
20 sophisticated than linear regression because it takes into account that measures within one
21 subject are correlated with repeated observations. An exchangeable correlation structure was
22 used for the GEE analysis in this study since it fit the data well (high within-subject correlations)
23 and for simplicity to minimize the number of parameters needed. The effect of surface type on
24 accelerometry-derived parameters was evaluated in three models: Firstly an unadjusted model,
25 the effect of surface type (woodchips and synthetic) compared to concrete (control reference
26 category) on each accelerometry outcome measure. Secondly, we assessed if training status

1 (highly-trained vs. recreational) was a confounding variable to the model, since trunk
2 accelerometry parameters have been found to significantly differ between trained and untrained
3 runners [21]. From this step, training status was only included in the model if it significantly
4 changed any of the regression coefficients for surface type (>10%). Thirdly, we assessed the
5 potential role of running speed as a confounder to surface type by adding it as a time-dependent
6 continuous covariate, since some trunk accelerometry parameters show strong relationships
7 with gait speed during running [21].

8 **Results**

9 Two subjects from the group of competitive runners were excluded from analysis since the
10 investigator deemed the attachment of their accelerometer to not be securely fasted upon
11 removal, and body sweat interfered with attempts at reattachment. Characteristics of the
12 remaining participants are shown in **Table 1**.

13 GEE results of the crude analysis for surface effects on accelerometry-derived parameters are
14 shown in **Fig. 1**. Synthetic track did not significantly change from concrete besides one dynamic
15 stability parameter (higher vertical stride regularity). Woodchip trail changed significantly from
16 concrete for several parameters, including spatio-temporal (lower step frequency), dynamic
17 stability (lower vertical but higher AP ratio of acceleration RMS, and lower step regularity and
18 stride regularity in the ML direction only) and dynamic loading (lower vertical and AP median
19 frequencies). The downward shift in vertical and AP median frequencies during stance from
20 concrete to woodchips can be observed in **Fig. 2**, overlaid with plots of trunk acceleration
21 signals during stance in the frequency domain.

22 When either training status (model two) or running speed (model three) were added as
23 potential confounders, statistical outcomes related to surface effects were unchanged. Therefore
24 all results pooled trained and untrained runners together (n = 28) and results from the crude
25 analysis on surface effects were reported (**Table 2**).

1 **Discussion**

2 Despite the frequently cited benefits of using wireless accelerometers to assess running gait in
3 ecological (i.e. real-world) rather than traditional (i.e. laboratory) settings, few studies have
4 actually done so. Therefore, this study sought to investigate outdoor surface effects on dynamic
5 stability and loading during running using tri-axial trunk accelerometry. Importantly, we found that
6 there were no significant ($p > 0.05$) differences in all parameters with exception to vertical
7 stride regularity between concrete and synthetic track. In contrast, woodchip trail altered
8 measures of dynamic stability compared to concrete; revealing significantly higher AP ratio of
9 acceleration RMS as well as lower ML inter-step and -stride regularity. Woodchip trail
10 additionally decreased median frequencies of both vertical (impact) and AP (breaking)
11 accelerations compared to concrete. In light of these results, it is reasonable to hypothesize that
12 differences may exist in injury risk and performance between concrete and woodchip running
13 surfaces.

14 In agreement with previous research [5,13,14], we found that contact time was unaffected by
15 running surface. On the other hand, we did find significantly reduced step frequencies on
16 woodchips. Thus, our hypothesis based on the principle of smoothness of CoM trajectory under
17 different surface conditions [5], active “self-stabilization” [8,9] or maintenance of global support
18 kinematics over different terrain [5,6] was not completely supported. From a biomechanical
19 perspective, woodchip trails differ fundamentally from concrete road and synthetic track due to
20 the presence of variously sized detached or scattered particles. Both compression and
21 displacement of the woodchips under the foot may then occur with each consecutive running
22 stride, causing dynamic instability and forcing lower-limb musculature to provide additional
23 work to the point of reaction force on the surface [26]. Therefore, the irregular nature of
24 woodchips could have interfered with the step length-step frequency relationship, as has
25 previously been observed when running on irregular [30] or rough [11] terrain.

1 The directional-shift in variation (ratio of acceleration RMS) from vertical to AP as well as the
2 decrease in inter-step and inter-stride regularity mediolaterally could indeed also be directly
3 related to the woodchip properties. This is consistent with past research, which has shown
4 similar destabilizations and directional shift from vertical to horizontal accelerations when
5 walking on uneven ground [31]. Thus, the runner's dynamic stability may be compromised as a
6 result of the irregularity of the uneven woodchips. Based on previous research in our laboratory
7 [20], running-related fatigue on an indoor treadmill results in a 13% increase in the AP ratio of
8 acceleration RMS. The runners in this study showed a 7% increase in the AP ratio of acceleration
9 RMS from concrete to woodchip surface. Thus, although both internal (fatigue) and external
10 (surface) factors contribute to destabilizing the stability of running, the magnitude of changes
11 due to running surface appear to be relatively smaller. It would be interesting to examine the
12 destabilizing running-related fatigue effects on a range of running surfaces, including
13 woodchips, since energy expenditure is increased when running on uneven terrain [8]. This
14 could provide insight into whether uneven surfaces such as woodchips are more detrimental to a
15 runner's stability when in a fatigued state.

16 Dynamic loading parameters were analyzed to provide information on magnitude and
17 proportion of propagated shock waves reaching the spine. We found no significant surface
18 effects for the amplitudes of vertical impact shock or AP breaking peak in the time domain. It is
19 possible that the amplitude of impact shock accelerations reaching the spine were unaffected
20 due to initial impact attenuations by the lower extremity, acting as a low-pass filter [32].
21 However, we also analyzed the frequency content of impact shock waves to gain better insights
22 in the distribution of the power content of impact, and observed a significant downward shift in
23 the median frequencies of vertical impact and AP breaking accelerations on woodchips. Visual
24 inspection of the frequency curves in **Figure 2** indicates that on woodchips a larger proportion
25 of both vertical and AP accelerations during stance were contained in the low frequency
26 component (4-8 Hz), while the proportion in the high frequency "impact" component (10-20 Hz)
27 appeared relatively unaltered. These results suggest that during stance a greater proportion of

1 accelerations may be needed for voluntary movements [29] and stability of the CoM on
2 woodchips, rather than any additional “structural dampening” provided as has anecdotally been
3 suggested.

4 Although we observed no confounding influence of training status, it is reasonable to argue that
5 maintaining dynamic stability could be more difficult for recreational compared to highly
6 trained runners [21]. Unfortunately, effect modification i.e. surface type x group interaction was
7 not directly investigated here due to low sample size and is a limitation of the current study.
8 Secondly, given that competitive runners were more familiar with synthetic track while
9 recreational runners were more familiar with wood chip trail, another limitation worth
10 mentioning is that surface habituation was unaccounted for. However, all participants had at
11 least some experience with running on all three surfaces and familiarization trials were
12 provided for each running surface to help minimize any immediate psychological adjustments.

13 We found that all surface effects were unaffected when running speed was added to the GEE
14 model as a covariate. The need to control for running speed was warranted given that previous
15 research has indicated that adjustments in running mechanics can often be explained by variable
16 running speed [21], presenting a major analytical problem. Additionally, running speed in itself
17 may be an adjustment to outdoor surfaces, even when pacing methods are enforced [11]. In
18 contrast, our in-field and statistical approach to deal with running speed as a potential
19 confounder enabled our runners to self-select speeds that were comfortable to them, with
20 arguably more real-world applicability.

21 **Conclusion**

22 The current results suggest that woodchip trails alter running mechanics by disrupting aspects
23 of dynamic stability and loading. The analysis presented here provides further insights into
24 running gait adaptations in typical, real-world settings.

1 References

- 2 [1] van Gent RN, Siem D, van Middelkoop M, van Os AG, Bierma-Zeinstra SMA, Koes BW.
3 Incidence and determinants of lower extremity running injuries in long distance runners:
4 a systematic review. *Br J Sports Med* 2007;41:469–80. doi:10.1136/bjsm.2006.033548.
- 5 [2] Zadpoor AA, Nikooyan AA. The relationship between lower-extremity stress fractures and
6 the ground reaction force: a systematic review. *Clin Biomech (Bristol, Avon)* 2011;26:23–
7 8. doi:10.1016/j.clinbiomech.2010.08.005.
- 8 [3] Milner CE, Ferber R, Pollard CD, Hamill J, Davis IS. Biomechanical factors associated with
9 tibial stress fracture in female runners. *Med Sci Sport Exerc* 2006;38:323–8.
10 doi:10.1249/01.mss.0000183477.75808.92.
- 11 [4] Taunton JE, Ryan MB, Clement DB, McKenzie DC, Lloyd-Smith DR, Zumbo BD. A
12 prospective study of running injuries: the Vancouver Sun Run “In Training” clinics. *Br J*
13 *Sports Med* 2003;37:239–44. doi:10.1136/bjsm.37.3.239.
- 14 [5] Ferris DP, Liang K, Farley CT. Runners adjust leg stiffness for their first step on a new
15 running surface. *J Biomech* 1999;32:787–94. doi:10.1016/S0021-9290(99)00078-0.
- 16 [6] Kerdok AE, Biewener AA, McMahon TA, Weyand PG, Herr HM. Energetics and mechanics
17 of human running on surfaces of different stiffnesses. *J Appl Physiol* 2002;92:469–78.
18 doi:10.1152/jappphysiol.01164.2000.
- 19 [7] Karamanidis K, Arampatzis A, Brüggemann GP. Adaptational phenomena and mechanical
20 responses during running: effect of surface, aging and task experience. *Eur J Appl Physiol*
21 2006;98:284–98. doi:10.1007/s00421-006-0277-7.
- 22 [8] Voloshina AS, Ferris DP. Biomechanics and energetics of running on uneven terrain. *J Exp*
23 *Biol* 2015;218:711–9. doi:10.1242/jeb.106518.
- 24 [9] Grimmer S, Ernst M, Gunther M, Blickhan R. Running on uneven ground: leg adjustment to
25 vertical steps and self-stability. *J Exp Biol* 2008;211:2989–3000. doi:10.1242/jeb.014357.
- 26 [10] Dixon SJ, Collop AC, Batt ME. Surface effects on ground reaction forces and lower
27 extremity kinematics in running. *Med Sci Sports Exerc* 2000;32:1919–26.
- 28 [11] Creagh U, Reilly T, Lees A. Kinematics of running on “off-road” terrain. *Ergonomics*
29 1998;41:1029–33.
- 30 [12] Tessutti V, Trombini-Souza F, Ribeiro AP, Nunes AL, Sacco IDCN. In-shoe plantar pressure
31 distribution during running on natural grass and asphalt in recreational runners. *J Sci*
32 *Med Sport* 2010;13:151–5. doi:10.1016/j.jsams.2008.07.008.
- 33 [13] Fu W, Fang Y, Liu DMS, Wang L, Ren S, Liu Y. Surface effects on in-shoe plantar pressure
34 and tibial impact during running. *J Sport Heal Sci* 2015;4:384–90.
35 doi:10.1016/j.jshs.2015.09.001.
- 36 [14] Tillman MD, Fiolkowski P, Bauer JA, Reisinger KD. In-shoe plantar measurements during
37 running on different surfaces: changes in temporal and kinetic parameters. *Sport Eng*
38 2002;5:121–8. doi:10.1046/j.1460-2687.2002.00101.x.
- 39 [15] Kobsar D, Olson C, Paranjape R, Hadjistavropoulos T, Barden JM. Evaluation of age-related
40 differences in the stride-to-stride fluctuations, regularity and symmetry of gait using a
41 waist-mounted tri-axial accelerometer. *Gait Posture* 2014;39:553–7.
42 doi:10.1016/j.gaitpost.2013.09.008.
- 43 [16] Tura A, Raggi M, Rocchi L, Cutti AG, Chiari L. Gait symmetry and regularity in transfemoral
44 amputees assessed by trunk accelerations. *J Neuroeng Rehabil* 2010;7:4.
45 doi:10.1186/1743-0003-7-4.
- 46 [17] Sekine M, Tamura T, Yoshida M, Suda Y, Kimura Y, Miyoshi H, et al. A gait abnormality

- 1 measure based on root mean square of trunk acceleration. *J Neuroeng Rehabil*
2 2013;10:118. doi:10.1186/1743-0003-10-118.
- 3 [18] Saether R, Helbostad JL, Adde L, Brændvik S, Lydersen S, Vik T. Gait characteristics in
4 children and adolescents with cerebral palsy assessed with a trunk-worn accelerometer.
5 *Res Dev Disabil* 2014;35:1773–81. doi:10.1016/j.ridd.2014.02.011.
- 6 [19] Le Bris R, Billat V, Auvinet B, Chaleil D, Hamard L, Barrey E. Effect of fatigue on stride
7 pattern continuously measured by an accelerometric gait recorder in middle distance
8 runners. *J Sports Med Phys Fitness* 2006;46:227–31.
- 9 [20] Schütte KH, Maas EA, Exadaktylos V, Berckmans D, Vanwanseele B. Wireless tri-axial
10 trunk accelerometry detects deviations in dynamic center of mass motion due to running-
11 induced fatigue. *PLoS One* 2015;1–12. doi:10.1371/journal.pone.0141957.
- 12 [21] McGregor SJ, Busa MA, Yaggie JA, Bollt EM. High resolution MEMS accelerometers to
13 estimate VO₂ and compare running mechanics between highly trained inter-collegiate
14 and untrained runners. *PLoS One* 2009;4:e7355. doi:10.1371/journal.pone.0007355.
- 15 [22] Kobsar D, Osis ST, Hettinga BA, Ferber R. Classification accuracy of a single tri-axial
16 accelerometer for training background and experience level in runners. *J Biomech*
17 2014;47:2508–11. doi:10.1016/j.jbiomech.2014.04.017.
- 18 [23] Radin EL, Orr RB, Kelman JL, Paul IL, Rose RM. Effect of prolonged walking on concrete on
19 the knees of sheep. *J Biomech* 1982;15:487–92. doi:10.1016/0021-9290(82)90002-1.
- 20 [24] Barrey E, Landjerit B, Wolter R. Shock and vibration during the hoof impact on different
21 track surfaces. *Equine Exerc Physiol* 1991;3:97–106.
- 22 [25] Moe-Nilssen R, Helbostad JL. Estimation of gait cycle characteristics by trunk
23 accelerometry. *J Biomech* 2004;37:121–6. doi:10.1016/S0021-9290(03)00233-1.
- 24 [26] Gaudino P, Gaudino C, Alberti G, Minetti AE. Biomechanics and predicted energetics of
25 sprinting on sand: Hints for soccer training. *J Sci Med Sport* 2013;16:271–5.
26 doi:10.1016/j.jsams.2012.07.003.
- 27 [27] Richman JS, Moorman JR. Physiological time-series analysis using approximate entropy
28 and sample entropy. *Am J Physiol Heart Circ Physiol* 2000;278:H2039–49.
- 29 [28] Voloshin AS, Mizrahi J, Verbitsky O, Isakov E. Dynamic loading on the human
30 musculoskeletal system - effect of fatigue. *Clin Biomech* 1998;13:515–20.
- 31 [29] Shorten MR, Winslow DS. Spectral analysis of impact shock during running. *Int J Sport*
32 *Biomech* 1992;8:288–303.
- 33 [30] Warren WH, Young DS, Lee DN. Visual control of step length during running over
34 irregular terrain. *J Exp Psychol Hum Percept Perform* 1986;12:259–66.
35 doi:10.1037/0096-1523.12.3.259.
- 36 [31] Menz HB, Lord SR, Fitzpatrick RC. Acceleration patterns of the head and pelvis when
37 walking on level and irregular surfaces. *Gait Posture* 2003;18:35–46.
- 38 [32] Verbitsky O, Mizrahi J, Voloshin A, Treiger J, Isakov E. Shock transmission and fatigue in
39 human running. *J Appl Biomech* 1998;14:300–11.

40

41 **Conflict of Interest Statement**

42 No conflicts of interest exist.

Table 1. Descriptive results of participant characteristics.

	Mean (SD)
n	28
Male (female)	14 (14)
Age (years) (SD)	22.62 (3.07)
Height (m) (SD)	1.76 (6.24)
Weight (kg) (SD)	63.05 (5.57)
Training volume (km·week ⁻¹) (SD)	41.22 (9.91)
Recreational (n = 17)	9.56 (11.88)*
Highly-trained (n = 11)	72.88 (7.94)
Concrete road running speed (m·s ⁻¹) (SD)†	3.79 (0.51)
Recreational (n = 17)	3.56 (0.44)*
Highly-trained (n = 11)	4.02 (0.58)
Synthetic track running speed (m·s ⁻¹) (SD)†	3.73 (0.45)
Recreational (n = 17)	3.54 (0.42)*
Highly-trained (n = 11)	3.92 (0.48)
Woodchip trail running speed (m·s ⁻¹) (SD)†	3.73 (0.45)
Recreational (n = 17)	3.47 (0.23)*
Highly-trained (n = 11)	3.99 (0.51)

†: based on self-selected speeds acquired from timing gates

*: t-test detected significantly different from highly-trained group ($p < 0.05$).

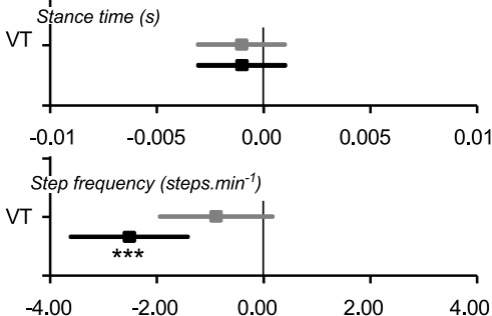
Table 2. Descriptive results (mean; SD) of accelerometry-derived parameters for repeated measures (n = 28).

Running gait parameter	Axis	Concrete road	Synthetic track	Woodchip trail
Spatio-temporal				
Step frequency (steps.min ⁻¹)	VT	169.75 (7.73)	169.03 (8.75)	167.4 (7.31)
Stance time (s)	VT	0.22 (0.02)	0.22 (0.02)	0.22 (0.02)
Dynamic stability				
Ratio of acceleration RMS (a.u)	VT	1.10 (0.08)	1.09 (0.07)	1.07 (0.07)
	ML	0.47 (0.11)	0.48 (0.11)	0.49 (0.11)
	AP	0.42 (0.10)	0.43 (0.10)	0.45 (0.09)
Step regularity (a.u)	VT	0.8 (0.09)	0.82 (0.08)	0.81 (0.08)
	ML	0.55 (0.13)	0.57 (0.12)	0.51 (0.12)
	AP	0.58 (0.12)	0.59 (0.13)	0.55 (0.11)
Stride regularity (a.u)	VT	0.81 (0.09)	0.84 (0.06)	0.82 (0.08)
	ML	0.69 (0.12)	0.70 (0.09)	0.64 (0.10)
	AP	0.65 (0.12)	0.67 (0.13)	0.63 (0.12)
Sample entropy (a.u)	VT	0.12 (0.02)	0.12 (0.02)	0.12 (0.02)
	ML	0.32 (0.07)	0.32 (0.07)	0.32 (0.07)
	AP	0.37 (0.10)	0.38 (0.11)	0.38 (0.11)
Dynamic loading				
Impact peak (g)	VT	4.02 (1.54)	3.91 (1.39)	3.67 (1.40)
Breaking peak (g)	AP	1.77 (0.60)	1.85 (0.70)	1.84 (0.66)
Median frequency during stance (Hz)	VT	16.19 (5.90)	14.82 (5.29)	13.90 (4.14)
	AP	15.55 (5.34)	14.99 (5.17)	13.72 (5.04)

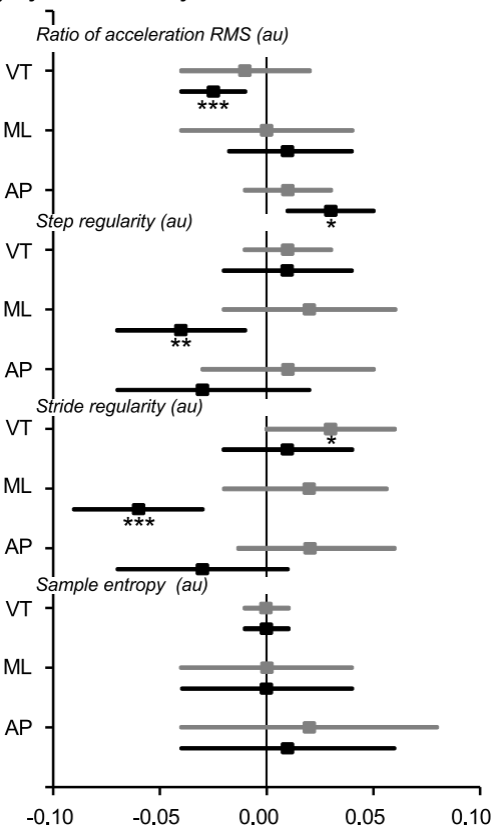
VT: vertical; ML: mediolateral; AP: anteroposterior; a.u: arbitrary units

A) Spatio-temporal

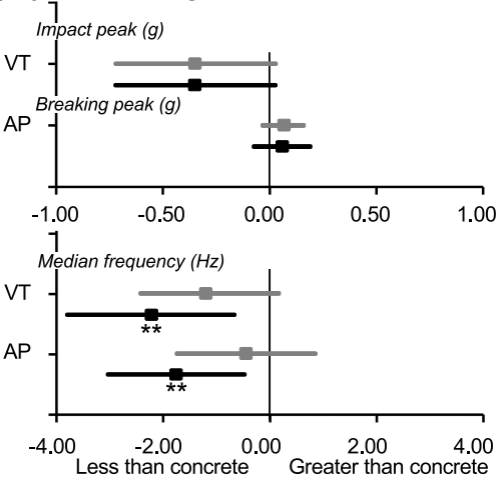
■ Mean change synthetic
■ Mean change woodchip
— 95% confidence interval



B) Dynamic stability



C) Dynamic loading



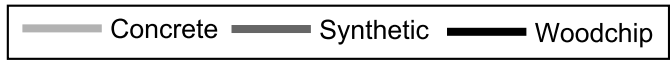
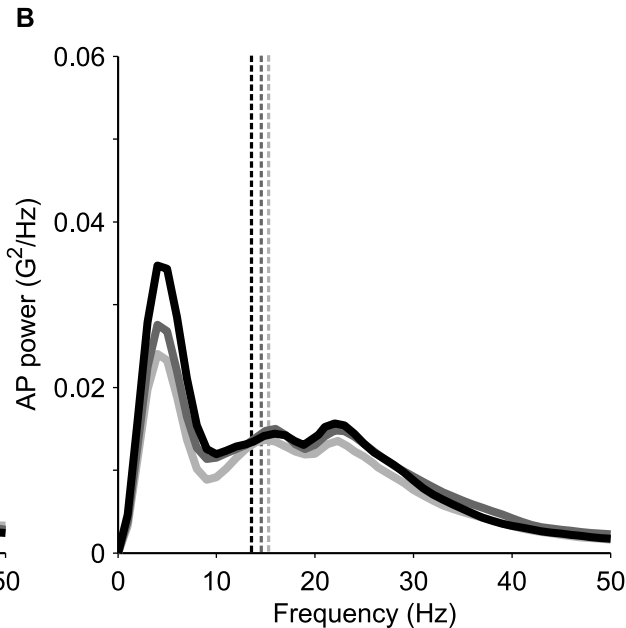
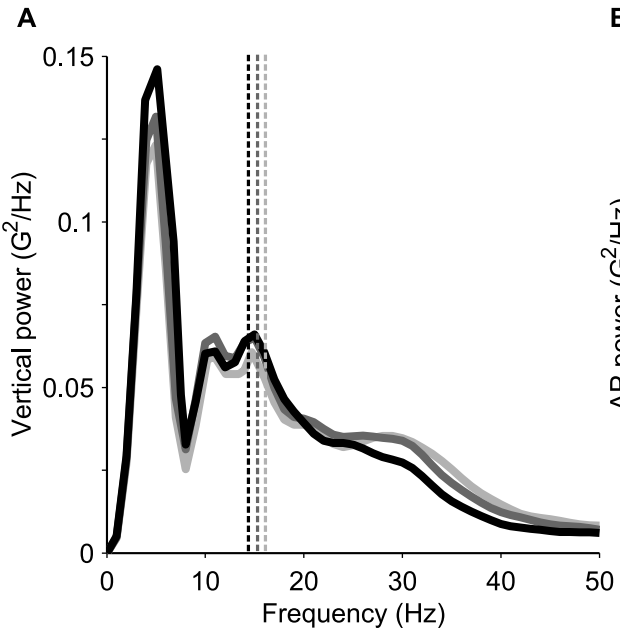


Fig. 1. Regression coefficients (95% confidence intervals) regarding GEE results for surface effects on accelerometry-derived parameters for repeated measures (n = 28). VT: vertical; ML: mediolateral; AP: anteroposterior; a.u: arbitrary units. */**/** Regression coefficient significantly different from reference category concrete road surface ($p < 0.05$)/ ($p < 0.01$)/ ($p < 0.001$).

Fig. 2. Group mean (n = 28) power spectra of A) vertical and B) AP trunk acceleration signals compared between concrete (light grey), synthetic track (dark grey), and woodchips (black). Vertical dashed lines indicate the median frequency for each surface respectively.