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Patterns of Gait Variability Across the Lifespan in Persons With and Without Down Syndrome

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**Using Nonlinear Measures to Understand Patterns of Gait Variability Across the Lifespan
in Persons with and without Down Syndrome.**

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7 the University of Michigan.

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1 **ABSTRACT**

2 **Background and Purpose:** Our aim here is to build upon the observation of higher
3 amounts of gait variability in persons with Down syndrome (DS) and describe the *patterns* of
4 that variability across the lifespan. Without knowing what baseline patterns look like and how
5 they relate to adaptive use of variability during gait it is difficult for physical therapists to
6 determine if and when to intervene and if increasing or decreasing variability is better. **Methods:**
7 We examined differences in patterns of gait variability in new walkers, preadolescents, and
8 adults with DS and typical development (TD) using the nonlinear measures of Lyapunov
9 Exponent (LyE) and Approximate Entropy (ApEn). Participants walked on a treadmill while we
10 collected 3-D motion analysis data. **Results/Discussion:** Within the higher amount of gait
11 variability persons with DS demonstrate across age compared to peers with TD, we found
12 significant differences in nonlinear measures of *patterns* of variability. Preadolescents
13 demonstrated higher LyE and ApEn values than new walkers and adults, suggesting they are
14 more adaptive in their use of variability during gait. **Clinical Interpretation/Conclusion:** From
15 a clinical perspective, our results suggest that physical therapists may focus interventions on
16 increasing adaptive use of variability during gait in new walkers and adults with DS. Experience
17 with increased variability through practice under variable conditions or with perturbations may
18 improve adaptive use of variability during gait.

1 **BACKGROUND AND PURPOSE**

2 People often demonstrate increased amount of variability in movement trajectories with
3 aging.¹⁻⁷ Consequently, a decrease in the amount of gait variability, finger force variability and
4 finger movement variability are all cited as positive outcomes of rehabilitation interventions for
5 older adults.⁸⁻¹¹ Although decreasing variability toward levels similar to younger persons can
6 have a positive effect, a full understanding of how variability relates to the control of movement
7 is still being discovered. Recently, scientists and clinicians have recognized that both the amount
8 and pattern of variability observed within movement trajectories over time can affect control of
9 movement. Impaired movement patterns can contain too much or too little variability, and
10 additionally the patterns of variability present within any amount of variability may contribute to
11 more or less adaptive control of movement.¹²⁻¹⁵ Thus both amount and pattern of variability
12 should be considered when trying to understand adaptive control of movement.

13 Down syndrome (DS) is one example of a population often described as demonstrating
14 increased amount of variability. Across the lifespan, persons with DS demonstrate more
15 variability in movement trajectories compared to their peers with typical development (TD).¹⁶⁻¹⁹
16 Persons with DS differ from persons with TD in some neurophysiologic and musculoskeletal
17 characteristics, including hypotonia, high ligamentous laxity and reduced capacity to produce
18 muscle force. We believe these conditions increase the challenge of dynamic upright posture,
19 especially in the earlier and later stages of life, and lead to the emergence of not only more
20 variable but also unique gait patterns.^{16-18,20} Although persons with DS demonstrate higher
21 amounts of variability in movements like walking and gripping compared to their peers with TD,
22 some research suggested that they use this variability functionally, to compensate for their
23 biomechanical instability, thus this variability level is optimal for them.^{21,22} If this is uniformly

1 true, then it might not be advisable for physical therapists to intervene with the intention of
2 decreasing the amount of variability in functional behaviors. One way to investigate this is to
3 study the patterns of variability within the higher amount of variability and relate them to
4 adaptive control of movement.

5 Current literature suggests that physical therapists should have their patients practice with
6 increased or decreased amounts of movement variability, as needed, to help them learn adaptive
7 use of variability.¹³⁻¹⁵ Without an understanding of what baseline amount and pattern of
8 variability looks like within a particular population and how it relates to functional control of
9 movement these are very difficult choices to make. In the case of adults with DS, we know they
10 demonstrate higher amounts of variability in gait and are more likely to have a history of falls
11 than their peers with TD.²³ Although increased amount of gait variability is related to increased
12 likelihood of falls and mobility disability in older adults with TD²⁴⁻²⁷ it is not clear whether a
13 causal link exists for adults with DS. Although adults with DS demonstrate increased amount of
14 gait variability, younger persons with DS also have high amounts of variability. It is possible that
15 adults have similar amounts and patterns of gait variability to preadolescents and other factors
16 contribute to their falls, or it is possible that adults have experienced changes in their ability to
17 adaptively use variability during gait and these changes negatively affect their gait patterns and
18 possibly link to an increased likelihood of falls. Our aim here is to expand our knowledge of
19 increased amounts of gait variability in persons with DS by describing their *patterns* of gait
20 variability across the lifespan. We will interpret our findings in relation to decisions physical
21 therapists need to make regarding efforts to affect gait variability in this population.

22 We examined changes in patterns of walking variability across the lifespan in persons
23 with and without DS using the nonlinear tools of Lyapunov Exponent (LyE) and Approximate

1 Entropy (ApEn). We used LyE to quantify the local stability (overlap or dispersion) of trajectories
2 of knee movement from one stride to the next. We used ApEn to quantify the regularity of the
3 patterns observed in size of successive step widths and step lengths. Previous work has shown
4 that 8-10-year-old children with DS had higher LyE and ApEn values indicating less local
5 stability and less regularity in their patterns of lower extremity segmental angles during walking
6 compared to their peers with TD.¹⁹ We hypothesized here that because preadolescents are at their
7 performance peak in terms of skill and efficiency, new walkers and adults with DS would show
8 less locally stable, less regular trajectories of movement (larger LyE and ApEn values) than
9 preadolescents with DS. Further, due to the inherent group differences in body structure and
10 function, we predicted that persons with DS would demonstrate less locally stable, less regular
11 trajectories of movement (larger LyE and ApEn values) across the lifespan, compared to their
12 peers with TD.

13 **METHODS**

14 **Data Collection**

15 Participants with DS and TD, representing three developmental levels: new walkers,
16 preadolescence and adulthood, came to the Developmental Neuromotor Control Laboratory at
17 the University of Michigan (total n= 58; Table 1). Participants were recruited through various
18 community activity and support groups in Michigan and Northern Ohio. They all participated in
19 adequately-powered studies with similar protocols in which gait measures (but not nonlinear
20 measures) were the primary dependent variables. The University of Michigan Institutional
21 Review Board approved all procedures. Prior to participation, we explained our study to
22 participants and caregivers. Participants signed an assent or consent form as appropriate, with
23 consent for assenting adults and children provided by legal guardians. Toddlers wore diapers

1 covered by black tights. Preadolescents and adults wore bathing suits or close-fitting shorts and
2 tank tops. We attached markers (2.5 cm diameter) bilaterally at the temporomandibular joint,
3 acromion process, lateral humeral epicondyle, styloid process, greater trochanter, femoral
4 condyle, 10 cm above lateral malleolus, heel bony prominence and third metatarsophalangeal
5 joint. We used a 6-camera Vicon Peak Motus* real-time system to collect 3-dimensional
6 reflective marker position data at a sampling rate of 60 Hz.

7 <<insert Table 1 approximately here>>

8 Participants walked barefoot over a 5.3-m GAITRite mat[†] 4-6 times at their preferred
9 speed. We used GAITRite software to calculate average walking speed for each participant and
10 subsequently determine belt speeds for treadmill trials.[‡] Based on previous work in our lab^{16,17},
11 we operationalized comfortable treadmill speed for all participants as 75% of their self-selected
12 overground speed. Comfortable speeds on a treadmill are slower than overground²⁸ and
13 participants with DS are cognitively not able to select their most comfortable speed in this novel
14 context. Participants performed two 30 s trials each at 45%, 75% and 110% of their overground
15 walking speed; trials progressed from slow to fast speeds. All participants walked without
16 touching the handrail and were guarded closely as they walked. Here we present results from the
17 75% speed only.

18 We used a Healthometer[§] scale to obtain body weight and a GPM anthropometer^{||} to
19 record height and body segment lengths. To assess motor task performance and developmental

* Vicon Peak Performance, 7388 South Revere Pkwy, Centennial, CO 80112.

† CIR Systems Inc, 60 Garlor Dr, Havertown, PA 19083.

‡ Parker brand, LET Medical Systems Corp., 5755 NW 151st Ave, Miami Lakes, FL 33014.

§ Precision Weighing Balances, 10 Peabody St, Bradford, MA 01835.

1 levels, we used age-appropriate instruments: the motor component of the Bayley Scales of Infant
2 Development[#] (new walkers); the 8-item balance subtest of the Bruininks-Oseretsky Test of
3 Motor Proficiency^{**} (preadolescents) and Berg Balance Scale^{29,30} (adults).

4 **Data Analysis: Theory and Definitions**

5 Stability, regularity and adaptability of gait can be defined in multiple ways. We use the
6 term stability here in reference to the LyE values, which quantify the local stability (overlap or
7 dispersion) of trajectories of movement from one repetition to the next. We use the term
8 regularity in reference to the patterns observed in size of successive step widths and step lengths,
9 as calculated by ApEn. We define the most adaptable gait patterns as those that are mid-range
10 (although not necessarily the middle) on the continuum of LyE and ApEn values.^{13,14,31} Mid-
11 range LyE values are considered adaptive as they represent patterns of variability that are neither
12 too stable (i.e., rigid) or too unstable, while mid-range ApEn values are considered adaptive as
13 they represent patterns of variability that are neither too regular (i.e., rigid) or too irregular.^{13,14}

14 We have not provided a tutorial on the use of nonlinear measures here because previous
15 publications have done so and discussed clinical applications.^{13,32,33} Briefly, LyE measures the
16 divergence within the trajectories of entire movement cycles, such as walking strides, by
17 quantifying their exponential separation in state space. We used LyE to measure the divergence

|| Siber Hegner and Co, Wiesenstr 8, PO Box 888, Zurich, Switzerland 8034.

The Psychological Corporation, San Antonio, TX 78283. 1993.

** American Guidance Service, 4201 Woodland Rd, Circle Pines, MN 55014.

†† Applied Nonlinear Sciences, LLC and Randle, Inc, Del Mar, CA 92014.

‡‡ MATLAB, The Mathworks Inc., 3 Apple Hill Dr., Natick, MA 01760.

§§ SPSS Inc., and IBM company, 233 S Wacker Dr., Chicago, IL 60606.

1 in the trajectory of the knee joint marker from one stride to the next. Figure 1 shows an example
2 of how we calculated LyE from the knee marker displacement in the vertical direction. Larger
3 values (closer to 0.5) indicate more dispersion, possible randomness and less similarity between
4 the trajectories of successive walking strides. Shifts toward smaller values (close to 0) indicate
5 less divergence, possible rigidity and more similarity between the trajectories of successive
6 walking strides.³⁴ ApEn quantifies the regularity of the pattern within a time series. ApEn values
7 exist on a continuum of 0 (completely regular pattern) to 2 (completely irregular, lack of
8 pattern).³⁴ A long stride alternating consistently with a short stride represents a more regular
9 pattern than a random series of unique stride lengths, although both behaviors would be
10 recognizable as a cyclic pattern of walking with similar values for mean and range of stride
11 length as traditionally calculated.

12 <<insert Figure 1 approximately here>>

13 We also would like to point out that LyE calculations are based on continuous kinematic
14 data, in this case of the knee marker trajectory throughout the stride, while ApEn calculations are
15 based on discrete spatial-temporal variables, here we used step length and step width. We made
16 these nonlinear tool choices deliberately; LyE allowed us to assess the stability of the knee
17 trajectory throughout and across continuous successive strides, while ApEn calculations allowed
18 us to assess the regularity in step length and width from one step to the next. These are specific
19 gait characteristics often reported for typical and atypical populations.

20 **Data Analysis: Procedures**

21 For the LyE analysis, we needed to identify a reflective marker to represent the cyclical
22 motion of each stride through space. Our pilot analyses showed that the knee marker provided
23 cleaner and more clearly cyclic data than the hip, ankle, heel and toe markers. We analyzed only

1 the anterior-posterior and vertical direction time series of the left knee 3-D data as lateral motion
2 of the knee is not a significant contributor to stride dynamics during walking. We analyzed
3 displacement of the marker, as opposed to joint angles or acceleration or other possible variables,
4 because we measured displacement directly and using a direct measurement as the basis for
5 nonlinear calculations is particularly important to minimize error due to the nature of the
6 calculation. Time series lengths for LyE calculation were 276 points for new walkers and 1800
7 points for preadolescents and adults. For toddlers, these points reflect 7 or 8 strides, the
8 maximum number they can produce continuously on a treadmill at this point in developmental
9 time. For preadolescents and adults these points represent approximately 24-39 strides [See
10 Smith, Stergiou and Ulrich for examples of time series and toe, knee and hip time series and
11 discussion of application of LyE to short new walker data sets³⁵]. Time series lengths for ApEn
12 calculation were 48-78 steps for preadolescents and adults and 14-16 continuous steps for new
13 walkers.

14 Once all data sets were cropped, as necessary, to the correct length we extracted knee and
15 heel marker data and calculated step width, step length and stride length. In Table 2 we provide
16 group means for the gait parameters for the walking strides used here to calculate LyE and
17 ApEn.

18 <<insert Table 2 approximately here>>

19 Next we determined the parameters and tested assumptions necessary for LyE and ApEn
20 calculations; these methods have been explained in depth previously.³⁴ We calculated time delay
21 and embedding dimension values using Tools for Dynamics software.^{††} We found an average

†† Applied Nonlinear Sciences, LLC and Randle, Inc, Del Mar, CA 92014.

‡‡ MATLAB, The Mathworks Inc., 3 Apple Hill Dr., Natick, MA 01760.

1 time delay of 3 and embedding dimensions of 8 for toddlers' knee time series and 5 for
2 preadolescents and adults. The increased number for toddlers reflects the increased 'noise'
3 present in their movements.³⁵ We tested our data for deterministic structure (mathematically
4 defined as a non-random structure) using a surrogate data comparison method and Chaos Data
5 Analyzer (CDA) software Professional Version.³⁶ We did not find significant differences
6 between LyE values for the original and surrogate anterior-posterior direction new walker LyE
7 data, and thus excluded these data from further analysis. Failed surrogation indicated that these
8 data, although they were collected during walking, were not mathematically definable as having
9 a periodic structure, again reflecting the increased 'noise' present in toddlers' movements.³⁵
10 Finally, we calculated LyE and ApEn. We used CDA software to calculate LyE data and custom
11 MATLAB^{††} programs for ApEn values. We calculated ApEn for the successive step lengths or
12 widths using ApEn input parameters of $m = 2$ and $r = 0.2$.³⁴

13 **Statistical Methods**

14 Statistics were calculated using an alpha level of 0.5 and SPSS software, version 17.0^{§§}.
15 In most cases, we used 2 (group) by 3 (age) ANOVA full factorial models with Bonferroni
16 corrections and follow-up tests for linear and quadratic trends. For the anterior-posterior LyE
17 data, because new walker data could not be included, we used a 2 (group) by 2 (age) ANOVA
18 with Bonferroni corrections and follow-up tests for linear trends only as a quadratic trend is not
19 possible with only two data points. We used linear and quadratic trend tests within the ANOVA
20 to assess the shape of change across the lifespan within each group, using one test to examine the
21 trend across the groups with DS and another to examine the groups with TD. A linear trend
22 would indicate increase or decrease in the measure across the lifespan, while a quadratic trend

§§ SPSS Inc., and IBM company, 233 S Wacker Dr., Chicago, IL 60606.

1 would indicate a “U” or inverted “U” shape across the lifespan. It is also possible to have
2 significant linear and quadratic trends simultaneously. In this case, it means the data increase or
3 decrease greatly and then flatten out, so that the “U” or inverted “U” (quadratic trend) is
4 significant and the data also show a significant overall increase or decrease over time (linear
5 trend).

6 **RESULTS**

7 *LyE: Local Stability of Limb Trajectories Across Successive Strides*

8 To test for differences in stability of knee trajectories in the vertical direction across
9 strides, we used a 2 (group) by 3 (age) ANOVA with vertical direction LyE values as the
10 dependent variable. The age effect was significant ($F[2, 46] = 17.53, p < 0.01$), while the group
11 effect and group by age interaction were not (see Figure 2a).

12 For follow-up analysis, we tested for linear and quadratic trends within the three age
13 groups. For the DS group, the quadratic trend test was significant ($p < 0.01$) while the linear
14 trend test was not. Pairwise comparisons revealed that preadolescents had higher vertical
15 direction LyE values than new walkers or adults ($p < 0.01$ for all). For the TD group, the
16 quadratic trend test was again significant ($p < 0.01$) while the linear trend test was not. Follow-up
17 pairwise comparisons revealed that preadolescents had higher vertical direction LyE values than
18 new walkers and adults ($p < 0.01$ for all).

19 We used a similar a 2 (group) by 2 (age) ANOVA for differences in LyE values in the
20 anterior-posterior direction. This analysis did not include the new walkers, whose anterior-
21 posterior direction data failed surrogation analysis. The group effect was significant ($F[1, 34] =$
22 $14.44, p < 0.01$), as was the age effect ($F[1, 34] = 18.92, p < 0.01$). The group by age interaction

1 was not significant (see Figure 2b). Inspection of means for the group effect revealed higher LyE
2 values in the anterior-posterior direction for the group with DS, while the age effect showed
3 higher LyE values in preadolescents compared to adults.

4 For follow-up analysis, we tested for a linear trend within the two age groups. We found
5 a significant linear trend ($p = 0.01$) for the DS group, reflecting lower anterior-posterior LyE
6 values in adults than preadolescents. Follow-up in the TD group also revealed a significant linear
7 trend ($p = 0.04$), again reflecting lower anterior-posterior LyE values in adults than
8 preadolescents.

9 <<insert Figure 2 approximately here>>

10 ***ApEn: Regularity of Pattern of Successive Step Lengths and Widths***

11 To test for differences in step length ApEn values, we used a 2 (group) by 3 (age)
12 ANOVA. The age effect was significant ($F[2, 49] = 9.37, p < 0.01$), while the group effect and
13 group by age interaction were not (see Figure 3a). Inspection of means revealed an inverted “U”
14 shape with highest values in preadolescents.

15 For follow-up analysis, we tested for linear and quadratic trends within the three age
16 groups. Follow-up analysis for the DS group showed significant linear ($p = 0.03$) and quadratic
17 ($p = 0.05$) trends. Pairwise comparisons revealed that the DS new walkers had significantly
18 smaller ApEn step length values than the DS preadolescent ($p = 0.01$) or adults ($p = 0.01$) while
19 the quadratic trend indicated higher values in the preadolescents than younger or older
20 participants. For the TD group, the quadratic trend was significant ($p = 0.02$) and the linear trend
21 was not, indicating higher values in preadolescents and lower values in older and younger
22 participants.

1 To test for differences in ApEn of step width values, we used a 2 (group) by 3 (age)
2 ANOVA. The age effect was significant ($F[2, 49] = 15.23, p < 0.01$), while the group effect and
3 group by age interaction were not (see Figure 3b). Inspection of the means revealed an inverted
4 “U” shape with highest values in preadolescents.

5 For follow-up analysis, we tested for linear and quadratic trends within the three age
6 groups. For the DS group, follow-up analyses showed significant linear and quadratic trends ($p <$
7 0.01 for both) indicating higher ApEn values for preadolescents, lower values for adults and
8 lowest values for new walkers. We obtained a significant quadratic trend ($p = 0.01$) but not linear
9 trend for the TD group, demonstrating higher ApEn values for preadolescents and lower values
10 for adults and new walkers.

11 <<insert Figure 3 approximately here>>

12 **DISCUSSION**

13 Within the higher amount of gait variability persons with DS consistently demonstrate
14 compared to their peers with TD across the lifespan, we show here that they also experience an
15 inverted “U”-shaped developmental trajectory in their *patterns* of variability across age. Overall,
16 our results of higher LyE values indicating less stability and higher ApEn values indicating less
17 regularity for preadolescents suggests they may be more adaptable in their gait patterns at this
18 point than as new walkers or by 35 years and beyond. That is, at either end of their years of gait
19 experience their patterns of variability are relatively more stable and regular, thus rendering their
20 gait potentially less adaptable to changes in task or environmental conditions. From a clinical
21 perspective, this suggests that physical therapists may be able to intervene to improve gait
22 performance, specifically patterns of variability related to stability and regularity, in new walkers
23 and adults with DS. Additionally, although amount of gait variability may not change with

1 intervention, LyE and ApEn values could be used to quantify changes in patterns of gait
2 variability in response to intervention. We provide here a baseline description of mean LyE and
3 ApEn values for patterns of stability and regularity of gait variability observed across the
4 lifespan in persons with DS.

5 Preadolescents with DS demonstrate closer to optimal walking performance compared to
6 their younger and older peers with DS; they are able to produce more continuous walking strides
7 than new walkers and prefer to walk faster than adults. Previous researchers have described
8 preadolescents with DS as being in one of the most consistent periods in their lives as they have
9 had at least 6 years of walking practice accompanied by steady physical growth.¹⁷ Because
10 preadolescents demonstrate closer to optimal walking performance than their younger and older
11 peers, our interpretation of the values obtained is that their LyE and ApEn values are also at their
12 peak and that they have learned to use their variability to adapt as well as possible during
13 locomotion. Thus, for future research and application, preadolescents' LyE and ApEn values
14 may represent the best possible values and their younger and older peers' lower values represent
15 walking patterns that are less adaptable as a result of *too much* stability and regularity in the
16 pattern. This is in contrast to our hypotheses that their younger and older peers would have
17 higher values representative of walking patterns that are less adaptable as a result of *too little*
18 stability and regularity in the pattern. Theoretically, extreme values may be on either side of the
19 ideal value, and understanding where they are should begin to influence physical therapy
20 interventions as we discover more. For example, a treadmill walking intervention at a constant
21 speed may decrease amount of gait variability and promote more stable and more regular
22 patterns of variability while walking at different and changing speeds may increase amount of
23 gait variability and promote less stable and less regular patterns of variability. Interventions

1 designed to promote practice with increased or decreased amounts and more or less stable and
2 regular patterns of movement variability, as needed, should help patients learn better adaptive
3 use of variability.

4 One can, however, only interpret LyE and ApEn values in a relative way, on a continuum
5 as compared to similar data collected and analyzed in the same manner. With the algorithms used
6 in the software we applied to our data the LyE values lie on a continuum from 0 to 0.5. A
7 periodic sine wave, with no divergence from one trajectory to the next, produces a LyE value of
8 0. A random signal, with maximal divergence, produces an LyE value of 0.5. Our results showed
9 LyE values ranging from approximately 0.15 to 0.20, indicating that divergence in participants'
10 knee trajectories was closer to the periodic end of the continuum. Occasions where participants
11 with DS had higher LyE values than their peers with TD reflected more divergence in their
12 movement trajectories. In our data, because both young and older groups were lower, ideal
13 values for LyE appeared to be around 0.20, indicating a more ideal balance between stability and
14 adaptability of performance. This value, however, would not necessarily be the ideal for a
15 different population or for any other gait parameter of interest, such as ankle or center of mass
16 motion.

17 ApEn values exist on a continuum of 0 to 2. Complete regularity of pattern produces an
18 ApEn value of 0, while complete irregularity and lack of a pattern is represented by an ApEn
19 value of 2. Our participants' ApEn values ranged from approximately 0.12 to 0.48, indicating
20 that step widths and lengths were closer to the regular pattern end of the continuum. This is not
21 unexpected, as consecutive walking steps represent a more periodic and regular pattern as
22 compared to variables such as center of pressure sway. For step length and width, preadolescents
23 demonstrated higher ApEn values than their older and younger peers, indicating less regularity of

1 pattern across successive steps and more adaptability of behavior. In our data, ideal values for
2 ApEn appeared to be around 0.48, a balance between regularity and adaptability of performance.
3 As with our LyE results, this is not an ideal value that would necessarily apply in a different
4 population or to any other variable of interest.

5 Our results here show that patterns of gait variability are different across the lifespan
6 within the group of persons with DS, and suggest that preadolescents in both groups are able to
7 use higher amounts of variability in an adaptive way compared to their younger and older peers.
8 The nonlinear measures, for the most part, reflected lifespan differences and did not reflect
9 overall differences between the DS and TD groups. The lack of group differences in ApEn and
10 vertical direction LyE values may be related to statistical analysis characteristics. The power to
11 statistically demonstrate differences between groups decreased when the much larger age effects
12 were included in the same analysis. In a study with similar dependent variables and only one age
13 group, 8-10-year olds, group differences in LyE and ApEn of lower extremity segmental angles
14 were observed. Results indicated statistically higher LyE and ApEn values indicating less
15 stability and less regularity in patterns of lower extremity segmental angles for children with DS
16 when compared to their peers with TD.¹⁹ Our group means in this study for vertical direction
17 LyE and ApEn step length are consistent in direction with results from this previous study.¹⁹ It is,
18 however, difficult to compare the actual LyE and ApEn values we obtained to those of other
19 studies, as different parameters used lead to different values of LyE and ApEn. Despite slightly
20 different analysis techniques and/or dependent variables, our data do appear to be in similar
21 range as results from other studies of treadmill walking in various populations. Buzzi and Ulrich
22 obtained LyE values ranging from 0.12 to 0.2 and ApEn values of 0.22 to 0.52 for lower
23 extremity segmental angles of children with DS and TD.¹⁹ Jordan and colleagues calculated LyE

1 values of around 0.1 at the ankle near the walk-run transition speed in healthy adult females.³⁷
2 Stergiou and colleagues obtained LyE values around 0.10-0.12 and ApEn near 0.20 to 0.26 from
3 the knee flexion/extension angle from participants with and ACL-deficient and contralaterally-
4 intact knee.^{38,39}

5 **CONCLUSIONS**

6 Overall our work suggests that, given their inherent neurophysiological constraints,
7 persons with DS control their gait in a way that is functional for them. The quality of this
8 solution, however, varies with age, as the stability and regularity of their patterns of gait
9 variability are different across the lifespan. Preadolescents are typically at a performance peak
10 for their lifespan, with the nonlinear measures LyE and ApEn suggesting that they are more
11 adaptable in their walking patterns than younger and older age groups. New walkers may have
12 difficulty adapting because they lack experience with this skill. By 35 years of age and beyond,
13 adaptability of adults' gait may diminish due to a decline in their amount of walking or a
14 hesitation to challenge themselves during locomotion. These results suggests that physical
15 therapists may be able to intervene to improve gait performance, specifically patterns of
16 variability related to stability and regularity, in new walkers and adults with DS. The strategy
17 might be to provide practice with less stable and less regular patterns of variability using
18 perturbations of speed or terrain to promote adaptive use of gait variability. This strategy
19 assumes that practice will lead to improved performance despite the presence of hypotonia, high
20 ligamentous laxity and reduced capacity to produce muscle force. Our findings hint improved
21 performance may be possible, though, as these factors are present across the lifespan yet
22 preadolescents with DS demonstrate different patterns of variability compared to their younger
23 and older peers.

1 **LIMITATIONS**

2 We are limited in our ability to make claims about how gait variability patterns change
3 for a person across the lifespan as our data are cross-sectional as opposed to longitudinal. Our
4 interpretation of gait as more adaptable is also limited as we used a treadmill to collect data and
5 did not test adaptability of gait to external manipulations. We do, however, believe that our
6 context is an appropriate paradigm to lay a foundation of data. Nonlinear analyses are better
7 applied to longer bouts of continuous walking and using a treadmill allowed us to obtain long,
8 continuous walking trials from preadolescents and adults. Toddlers, however, are only able to
9 produce 7 or 8 continuous walking strides on a treadmill at this point in developmental time,
10 leading to shorter data sets than are typically used when performing nonlinear analyses. For these
11 reasons, our ability to apply nonlinear analyses to new walker data was limited [See Smith,
12 Stergiou and Ulrich for discussion of potential confounding effect of application of LyE to short
13 new walker data sets³⁵]. We also appreciate the need for software development to allow
14 clinicians to collect and analyze data without using research laboratory resources and for
15 correlation of nonlinear measures with clinical rating scales of body structure and function,
16 activity and participation.

1 **FIGURE CAPTIONS**

2

3 Figure 1. Lyapunov Exponent (LyE) calculation visual analogy, using data from consecutive
4 stride cycles of the knee marker of an adult participant with Down syndrome (DS). 1a is the knee
5 marker vertical position time series, 1b shows 3 strides extracted from the time series (1a) and
6 overlaid and 1c demonstrates a magnified version of an isolated segment of the state space to
7 show the divergence between neighboring trajectories.

8

9 Figure 2. Vertical (2a) and anterior-posterior (2b) direction Lyapunov Exponent (LyE) values for
10 participants with Down syndrome (DS) and typical development (TD). Age groups are as
11 follows: NW = new walkers, PA = preadolescents, A = adults.

12

13 Figure 3. Approximate Entropy (ApEn) of step length (3a) and step width (3b) for participants
14 with Down syndrome (DS) and typical development (TD). Age groups are as follows: NW =
15 new walkers, PA = preadolescents, A = adults.

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