Variability of Graphomotor Task Performance in Adults with ADHD: A Kinematic Approach

Thomas Duda

Follow this and additional works at: https://scholar.uwindsor.ca/etd

Recommended Citation
https://scholar.uwindsor.ca/etd/4802
Variability of Graphomotor Task Performance in Adults with ADHD: A Kinematic Approach

by

Thomas A. Duda

A Thesis
Submitted to the Faculty of Graduate Studies through Psychology
in Partial Fulfillment of the Requirements for the Degree of Master of Arts at the University of Windsor

Windsor, Ontario, Canada

2012

© 2012 Thomas A. Duda
Variability of Graphomotor Task Performance in Adults with ADHD: A Kinematic Approach

by

Thomas A. Duda

APPROVED BY:

______________________________________________
Dr. Nancy McNevin
Faculty of Human Kinetics

______________________________________________
Dr. Anne Baird
Department of Psychology

______________________________________________
Dr. Joseph Casey, Advisor
Department of Psychology

______________________________________________
Dr. Josee Jarry, Chair of Defense
Department of Psychology

September 13, 2012
DECLARATION OF ORIGINALITY

I hereby certify that I am the sole author of this thesis and that no part of this thesis has been published or submitted for publication.

I certify that, to the best of my knowledge, my thesis does not infringe upon anyone’s copyright nor violate any proprietary rights and that any ideas, techniques, quotations, or any other material from the work of other people included in my thesis, published or otherwise, are fully acknowledged in accordance with the standard referencing practices. Furthermore, to the extent that I have included copyrighted material that surpasses the bounds of fair dealing within the meaning of the Canada Copyright Act, I certify that I have obtained a written permission from the copyright owner(s) to include such material(s) in my thesis and have included copies of such copyright clearances to my appendix.

I declare that this is a true copy of my thesis, including any final revisions, as approved by my thesis committee and the Graduate Studies office, and that this thesis has not been submitted for a higher degree to any other University or Institution.
ABSTRACT

Attention-Deficit Hyperactivity Disorder (ADHD) is a neurodevelopmental disorder characterized by symptoms of inattention and/or a combination of hyperactivity and impulsivity. Motor problems, including poor graphomotor skills, are frequently found in those with ADHD and have been noted to be undertreated. Variability of performance within several domains has also been indicated as a hallmark of ADHD. The present study sought to 1) determine whether the variability of performance observed in other psychological domains in those diagnosed with ADHD manifests within kinematic variables of graphomotor output and 2) determine whether a novel writing task differentially affects the graphomotor output of adults diagnosed with ADHD versus controls. Findings and implications are discussed.

Keywords: digitizing tablet, stimulant medication, fine motor skills, variability of task performance
DEDICATION

To my wife Laura, without whose love and support, my current path would be unbearable.
The successful completion of this document would not have been possible without the support of several key individuals. I would like to thank my supervisor, Dr. Joseph Casey, for his advice, feedback, and guidance in managing this complex project. I would also like to acknowledge my thesis committee members, Dr. Anne Baird and Dr. Nancy McNevin, for all of their input and support throughout this process.

I would also like to express my appreciation to Vilija Petrauskas for completing fundamental groundwork to acquire the hardware and software necessary to conduct this study. Thank you as well to my fellow graduate students who have provided emotional support as well as advice concerning statistical analyses and the pragmatics of the study. Finally, I also thank my family – especially my wife Laura – for all of their patience and support throughout the past two years. I would not have been able to pursue and successfully complete the Masters portion of my training without them.
# TABLE OF CONTENTS

DECLARATION OF ORIGINALITY ................................................................. iii

ABSTRACT ........................................................................................................ iv

DEDICATION ...................................................................................................... v

ACKNOWLEDGEMENTS .................................................................................... vi

LIST OF TABLES ................................................................................................ ix

LIST OF FIGURES .............................................................................................. x

CHAPTER

I. INTRODUCTION

   Introduction ........................................................................................................ 1

II. REVIEW OF LITERATURE

   Diagnostic Criteria .............................................................................................. 2
   Etiology of ADHD ............................................................................................... 3
   Impairments Associated with ADHD ................................................................. 7
   Kinematic Analysis of Handwriting ..................................................................... 9
   Clinical Research Utilizing Kinematic Analysis ............................................... 12
   The Present Study ............................................................................................... 16

III. DESIGN AND METHODOLOGY

   Participants .......................................................................................................... 18
   Materials and Apparatus .................................................................................... 19
   Demographic Information and ADHD Symptomatology ............................... 19
   Estimate of Intellectual Ability ......................................................................... 21
   Kinematic Analysis and Digitizing Tablet ....................................................... 21
   Procedures .......................................................................................................... 23

IV. ANALYSIS OF RESULTS

   Data Analysis of Assumptions ......................................................................... 26
   Homogeneity of Variance ................................................................................... 26
   Normality and Independence of Observations ............................................... 29
   Demographic Information and ADHD Symptomatology ............................... 32
   Kinematic Analyses ............................................................................................ 36
   Overlap with Past Research ............................................................................. 36
Variability of Graphomotor Performance ........................................38
The Effects of Novelty on Variability ...........................................39

V. CONCLUSIONS AND RECOMMENDATIONS

Discussion .....................................................................................42
Overlap with Past Research .........................................................42
Variability of Graphomotor Performance .....................................45
Methodological Limitations .........................................................49
Future Research ...........................................................................50

APPENDICES

Appendix A: Interview Questionnaire ...........................................52
Appendix B: Scaled Version of Novel Symbol .................................54

REFERENCES ..................................................................................55

VITA AUCTORIS ............................................................................75
LIST OF TABLES

Table 1. Internal Consistency & Test-Retest Reliability of the BAARS-IV ........21

Table 2. Variance of Data Within Conditions ........................................28

Table 3. Levene’s Test of Homogeneity of Variance ...................................29

Table 4. Normality of Data .........................................................................31

Table 5. Participant Demographic Information ...........................................34

Table 6. BAARS-IV ADHD Symptomatology ...........................................35

Table 7. Number of Participants with BAARS-IV Scores ≥ 93rd %tile ..........35

Table 8. Automatized Writing Fluency – Summary and Source Table ........37

Table 9. Automatized Writing Variability – Summary and Source Table ....39

Table 10. Graphomotor Variability as a Function of Novelty.....................40
LIST OF FIGURES

Figure 1. Velocity profile of the word “hello” written fluently.................................14

Figure 2. Velocity profile of the word “hello” written with simulated dysfluency ....15

Figure 3. Automatized Writing Fluency of Control & Clinical Participants.............38

Figure 4. Automatized Writing Variability...............................................................39

Figure 5. Graphomotor Variability as a Function of Novelty.................................41
Attention-Deficit Hyperactivity Disorder (ADHD) is a neurodevelopmental disorder characterized primarily by symptoms of inattention and/or a combination of hyperactivity and impulsivity (Barkley, 2006). In addition to the diagnostic criteria that define ADHD, several other impairments have been consistently identified in those with ADHD. These characteristics include motor skill impairments, such as poor handwriting, and variability of task performance, which manifests within several domains. One promising method that has been used to investigate graphomotor functioning (i.e., handwriting) is kinematic analysis, which has historically involved the use of digitizing technology. Kinematic analysis of graphomotor functioning in the ADHD population has indicated that within the context of medication status (i.e., whether taking prescribed dosages of stimulant medication or having discontinued medication), children with ADHD differ in automatized graphomotor fluency when compared to unaffected children. Similar results have not been documented in adults with ADHD. However, no study has investigated whether the variability of performance that is observed within the ADHD population extends into the graphomotor domain.
CHAPTER II
REVIEW OF LITERATURE

Diagnostic Criteria

The most recent version of the Diagnostic and Statistical Manual of Mental Disorders (DSM-IV-TR; American Psychiatric Association, 2000) estimates the prevalence of ADHD to be between 3% and 7% of school aged children in the United States. In adults, the prevalence of ADHD has been estimated at approximately 4% (as cited in Biederman, 2005). Data demonstrating persistence of ADHD symptomatology from childhood into adulthood are mixed, with estimates ranging between 4% (Mannuzza, Klein, Bessler, Malloy, & LaPadula, 1998) and 85% (Barkley, Fischer, Smallish, & Fletcher, 2002). Although estimations may be conservative in general (Root & Resnick, 2003), Barkley (2006) has indicated that prevalence estimates of ADHD differs based on a variety of factors, including sex, age, diagnostic criteria, data collection methods, and country of origin.

Utilizing criteria described in the DSM-IV-TR, a diagnosis of ADHD can be given to those who demonstrate either “six (or more)” symptoms of inattention and/or “six (or more)” symptoms related to hyperactivity and impulsivity that “have persisted for at least 6 months to a degree that is maladaptive and inconsistent with developmental level” (American Psychiatric Association, 2000, p. 92). Further, these symptoms must have been observed before the individual was 7 years old and with impairment occurring in two or more settings (e.g., at school, in the home, and/or in the work-place). Specific subtypes of ADHD, which correspond to different combinations of symptomatology, include ADHD combined type (ADHD-C), ADHD Predominantly Inattentive Type
(ADHD-PI), and ADHD Predominantly Hyperactive-Impulsive Type (ADHD-HI). A diagnosis of ADHD-C requires that both six or more symptoms of inattention and six or more symptoms of hyperactivity-impulsivity have been present for at least the past six months. The ADHD-PI subtype is indicated when six or more symptoms of inattention are present for at least six months, but fewer than six symptoms of hyperactivity-impulsivity are present during this same time period. Finally, a diagnosis of ADHD-HI is appropriate if six or more symptoms related to hyperactivity-impulsivity have been present for at least the past six months, but fewer than six symptoms of inattention are present during this same time period.

**Etiology of ADHD**

The etiology of ADHD is complex in nature, although recent research implicates neurological and genetic factors as primary agents of pathogenesis (Barkley, 2006). The advent and subsequent popularity of modern neuroimaging techniques such as magnetic resonance imaging (MRI) and diffusion tensor imaging (DTI) have enhanced the ability of researchers to analyze the structural neuroanatomy of individuals in a non-invasive manner. In the case of ADHD, several consistent findings have emerged with regard to abnormal structure of the central nervous system (Barkley, 2006). Widespread reductions of cortical gray matter have been found in the frontal, parietal, temporal, and occipital lobes of the cerebral cortex in general (Batty et al., 2010; Shaw et al., 2006) and in frontal and posterior association cortices in particular (Narr et al., 2009). Although findings vary to some degree between studies, reductions in gray matter volume have been found in more circumscribed areas of the cortex and subcortical nuclei in both children and adults with ADHD. These areas include the prefrontal and dorsolateral
prefrontal cortices, basal ganglia (caudate nucleus, globus pallidus, putamen, and substantia nigra), and anterior cingulate cortex (Amico, Stauber, Koutsouleris, & Frodl, 2010; Castellanos, Geidd, Marsh, & Hamburger, 1996; McAlonan et al., 2007; Romanos et al., 2010; Seidman et al., 2011). Reductions in the infratentorial structural volume of the cerebellar vermis have also been found in both children (Castellanos et al., 2001; Durston et al., 2004; Mackie et al., 2007) and adults (Seidman et al., 2011) diagnosed with ADHD.

Research also indicates that although those diagnosed with ADHD do not consistently demonstrate global reductions in white matter volume compared to controls (Amico et al., 2010; Batty et al., 2010; Durston et al., 2004; McAlonan et al., 2007; Narr et al., 2009), reduced white matter volumes in specific areas of the cerebrum have been more consistently documented. For example, McAlonan et al. (2007) found that white matter tracts of the corpus callosum evidenced reduced volume in those diagnosed with ADHD. This finding is consistent with past studies indicating reduction in white matter of the corpus callosum in general (Hynd et al., 1991) and the splenium of the corpus callosum in particular (Semrud-Clikeman et al., 1994). Other studies investigating the structural integrity of white matter pathways connecting different regions of the cerebrum suggest that these pathways appear to be compromised in the ADHD population (Konrad & Eickhoff, 2010). More specifically, the superior longitudinal fasciculus and anterior corona radiata, which are tracts projecting between the frontal cortex and basal ganglia, have evinced reduced white matter integrity in children and adults based on measurements of fractional anisotropy (FA; representing the directionality and shape of the water molecules within the tract), mean diffusivity (MD), and apparent diffusion
coefficient (ADC; representing the volume of white matter diffusion) (Liston, Cohen, Teslovich, Levenson, & Casey, 2011). Due to the aforementioned inconsistencies in white matter volumetric findings in the ADHD literature, however, firm conclusions concerning the role of white matter pathways in the pathophysiology of ADHD cannot be drawn at this time.

Although relationships between ADHD symptomatology and structural abnormalities can only be inferred due to the nature of these studies, functional neuroimaging techniques, such as functional magnetic resonance imaging (fMRI), provide additional evidence that structures believed to subserve abilities related to attention, inhibition, and motor control – abilities that are impaired in those with ADHD – are the same structures that demonstrate structural abnormalities in ADHD (Brossard-Racine, Majnemer, & Shevell, 2011; Seidman et al., 2006; Swanson, Castellanos, Murias, LaHoste, & Kennedy, 1998; Shaw et al., 2006). Compared with healthy children, children with ADHD show abnormal patterns of activation (i.e., hypo-activation) in the prefrontal cortex, basal ganglia, and cerebellum when performing tasks related to attention, inhibition, motor control, and executive function (Bush et al., 1999; Durston et al., 2003; Posner et al., 2011; Rubia et al., 1999; Teicher et al., 2000; Vaidya et al., 1998; Yeo et al., 2003). Differences also appear to persist into adulthood. For example, Cubillo, Halari, Giampietro, Taylor, & Rubia (2011) found that compared with neurotypical individuals, medication naive adults diagnosed with ADHD in childhood who continued to demonstrate symptomatology into adulthood were found to have reduced activation in the orbital frontal cortex, medial frontal cortex, and striatum (i.e., basal ganglia) during tasks requiring inhibition, as well as reduced activation in the lateral
inferior and dorsolateral prefrontal cortices during tasks of working memory and attention.

Although studies have not demonstrated that ADHD occurs as a result of chromosomal abnormalities, several lines of research (i.e., family, adoption, twin, and genetic studies) indicate that ADHD has a high degree of heritability (Barkley, 2006). Highlighting the heritable and familial nature of ADHD – with some heritability rates estimated to be as high 0.76 (Faraone et al., 2005) – are findings that asymptomatic siblings of those diagnosed with ADHD demonstrate a trend towards similar volumetric reductions in cortical regions comparable with those found in their affected siblings (Durston et al., 2004).

At least seven genes appear to be implicated in the etiology of ADHD (Faraone et al., 2005), although several other genes are currently under investigation (see Banaschewski, Becker, Scherag, Franke, & Coghill, 2010, for a review). One example is the dopamine transporter gene, DAT1, which has received significant attention as mutations of this gene have been found to be related to presence of ADHD symptomatology in both adults (Brown et al., 2011) and children (Daly, Hawi, Fitzgerald, & Gill, 1999).

Dysfunction or imbalance of dopamine (DA), norepinephrine (NE), and noradrenaline (NA) neurotransmitters have also been implicated in the pathophysiology of ADHD (Arnsten, Berridge, & McCracken, 2009; Barkley, 2006; Biderman, 2005). In a recent review of the literature investigating the influences of DA and NA in ADHD, del Campo, Chamberlain, Sahakian, and Robbins (2011) suggested that DA and NA may play more specific roles in the presentation of ADHD symptomatology. That is, whereas
a combination of DA and NA abnormalities may affect functioning of the prefrontal cortex and by extension abilities related to inhibition, DA alone may affect functioning of subcortical structures such as the basal ganglia, in turn affecting attentional abilities.

**Impairments Associated with ADHD**

Beyond the primary symptoms of inattention, hyperactivity, and impulsivity, decades of research has demonstrated that ADHD is associated with numerous impairments affecting various domains of activities and functioning. Areas of impairment include cognitive functioning, language development and expression, motor skills, emotional regulation, academic performance, consistency of task performance, and general health and well-being (Barkley, 2006). Of particular interest here are the motor control problems, which are often under-treated in this population (Fliers et al., 2009), and the variability of task performance and expression.

Individuals diagnosed with ADHD have been shown to demonstrate variability in task performance and behaviours within several domains, including emotional expression (i.e., emotional lability; Barkley & Fischer, 2010; Posner et al., 2011), qualitative and quantitative handwriting production (Rosenblum, Epsztein, & Josman, 2008), in-phase bimanual coordination (Klimkeit, Sheppard, Lee, & Bradshaw, 2004), motor force output (Pereira, Eliasson, & Forssberg, 2000), and fine motor skill movements (Pitcher, Piek, & Barrett, 2002). Anecdotal reports from teachers and parents also suggest that children diagnosed with ADHD, as compared to healthy children, display a great deal of variability in their academic work and the quality with which they complete household duties (Barkley, 2006). Due to the observation that variability of task performance has been documented in several domains and appears to be ubiquitous in this population,
“some believe [variability of task performance] to be a primary deficit in ADHD” (Barkley, 2006, p. 136).

Although it is still unclear whether or not developmental motor milestones are generally delayed in children with ADHD (Barkley, 2006), the pervasive nature of motor difficulties that are observed in this population is highlighted by findings demonstrating significant comorbidity with Developmental Coordination Disorder, which is characterized by “marked impairment in the development of motor coordination” that “significantly interferes with academic achievement or activities of daily living” (American Psychiatric Association, 2000, pp. 56-57), when compared to the general population (Kadesjo & Gillberg, 2001; Piek, Pitcher, & Hay, 1999). Indeed, there is some evidence to suggest that ADHD and Developmental Coordination Disorder (DCD) share a genetic component (Martin, Piek, & Hay, 2006). Regardless of the presence of DCD, it is clear that those diagnosed with ADHD demonstrate motor impairments more frequently than the general population (Brossard-Racine, Majnemer, & Shevell., 2011). Examples of motor impairments found in those diagnosed with ADHD include poor handwriting (Brossard-Racine, Majnemer, Shevell, Snider, & Belanger, 2011); decreased speed and accuracy of complex (but not simple) fine and tactual motor performance (Meyer & Sagvolden, 2006); and deficits in balance, manual dexterity, coordination, and fine and gross motor skills (Piek, Pitcher, & Hay, 1999). Highlighting the importance of impairments in both motor functioning in general and timing of motor behaviour in particular are studies indicating that these problems are not only found in those diagnosed with ADHD, but also in siblings without an ADHD diagnosis. For example, Rommelse and colleagues (2008) found this relationship between affected and non-ADHD siblings.
and concluded that based on the evidence, “Variability in motor timing appears a useful endophenotypic candidate: It is clearly associated with ADHD, it is also present in non-ADHD siblings, and it correlates within families” (p. 131). “Moderate” and statistically significant positive correlations between severity of ADHD symptomatology and severity of motor sequelae have also been documented (Rommelse et al., 2009), which provide additional support for the notion that both motor control dysfunction and variability in task performance could be considered as primary deficits in those diagnosed with ADHD.

Relevant to the academic success of children is the skill of handwriting. In a review of the literature investigating the handwriting skills of children diagnosed with ADHD, Brossard-Racine, Majnemer, Shevell, and Snider (2008) concluded that the handwriting of individuals in this population can be characterized as impaired, often illegible, and less organized than the handwriting of control children, which in turn results in low academic achievement. Poor qualitative writing observed in this population does not appear to be related to pure visual-perceptual, visual-motor integration, or linguistic difficulties; instead, poor performance likely involves many different processes (Brossard-Racine et al., 2008), including dysfunction in basic parameter setting, such as regulation of force, speed, and size of graphomotor movements (van Galen, 1991); motor control; and timing aspects of handwriting (Adi-Japha et al., 2007; Marcotte & Stern, 1997; Schoemaker, Ketelaars, van Zonneveld, Minderaa, & Mulder, 2005).

**Kinematic Analysis of Handwriting**

The volitional control of handwriting can be thought of as a complex process involving the integration of “cognitive, psychomotor, and biophysical processes” (van
that are organized hierarchically and in parallel (Plamondon, 1995a) to produce meaningful visual-spatial output. Using a motor program metaphor, graphomotor processes begin with the retrieval of a high-level representation of the desired motor output - which might involve acquiring trajectory based stroke segments that can be combined to form complex symbols as opposed to retrieving whole letters or words stored within a visual-spatial “brain dictionary” (Lacquaniti, 1989, p. 287). This in turn is followed by a conversion of this representation into motor control “commands,” finally ending with the neuromuscular system responding in the desired manner (Plamondon, Yu, Stelmach, & Clement, 1991). In addition, the neuromuscular and higher-order systems make necessary adjustments based on relevant “visual and/or kinesthetic feedback” (Dooijes, 1983, p. 104). Central nervous system structures likely involved in these motor output processes include the primary motor cortex, premotor cortex, supplemental motor area, basal ganglia, cerebellum, and spinal cord (Plamondon, 1995a).

Studies investigating the cognitive, psychomotor, and biophysical processes involved in graphomotor control generally support this process and its related components (Meulenbroek & Thomassen, 1993; Meulenbroek & van Galen, 1988; Portier & van Galen, 1992; Teulings, Thomassen, & van Galen, 1983; van Galen, 1990; Woch, Plamondon, & O’Reilly, 2011; see Plamondon & Maarse, 1989, for a review and evaluation of computational motor models of handwriting), with the initial phases of voluntary motor control represented by measurements of reaction time and the latter phases represented by measurements of total movement time (Bellgrove et al., 1997) and other variables. The use of objective tools and methods to assess handwriting movements
(e.g., kinematic analysis), then, can be viewed as a method to make inferences about these cognitive, psychomotor, and biophysical processes underlying graphomotor function.

Kinematic analysis involves the quantification of “time changes of position, velocity, and acceleration” (Viviani & Terzuolo, 1982, p. 431). Although many technological options are available for kinematic analysis, the use of digitizing tablets to capture handwriting signals has predominated in graphonomic research of both healthy and clinical populations over the past 30 years (for a review of early graphomotor research, including the use of digitizing tablets, see Graham & Weintraub, 1996). In the domain of graphonomics, kinematic measures can be quantified using parameters of time, acceleration, velocity, and pen pressure, and variables derived from these basic measures can be used to (a) describe abilities related to degree of movement automatization and fluency (Eichhorn et al., 1996; Margolin & Wing, 1983; Mergl, Tigges, Schroter, Moller, & Hegerl, 1999; Portier & van Galen, 1992; Teulings, Contreras-Vidal, Stelmach, & Adler, 1997; Yan, Rountree, Massman, Doody, & Li, 2008); (b) quantify the relative decelerations and accelerations of handwriting movements (Eichhorn et al., 1996; Mergl et al., 1999; Plamondon & Clement, 1991; van Galen, Portier, Smits-Engelsman, & Schomaker, 1993); (c) indicate stability, coordination, and consistency of an individual’s handwriting (Mergl et al., 1999; Schroter et al., 2003; Teulings & Schomaker, 1993; Slavin, Phillips, Bradshaw, Hall, & Presnell, 1999); (d) indicate the sharing of processing resources, the difficulty of writing trajectories, and the presence of dysmetria (van Galen, 1991; Teulings, Contreras-Vidal, Stelmach, & Adler, 1997; Phillips et al., 2009); (e) quantify fine motor hypotonia and general proficiency (Mergl et al., 1999; Wann &
Nimmo-Smith, 1991; Phillips et al., 1999); and (f) indicate the smoothness and efficiency of movements (Bellgrove et al., 1997; Phillips et al., 2009). In this sense, the metrics produced by kinematic analyses of handwriting can be viewed as objective rather than subjective measurements of graphomotor performance.

**Clinical Research Utilizing Kinematic Analysis**

The use of digitizing technology to quantify graphomotor processes as an investigative and potentially diagnostic tool has been conducted with a multitude of patient populations. Pathologies and disorders investigated include, but are not limited to, ADHD (e.g., Adi-Japha et al., 2007; Flapper, Houwen, & Schoemaker, 2006; Schoemaker et al., 2005; Tucha & Lange, 2001, 2004, 2005; Tucha, Paul, & Lange, 2003); Dementia, Alzheimer’s Disease, and Mild Cognitive Impairment (e.g., Bellgrove et al., 1997; Yan et al., 2008); DCD (e.g., Bo, Bastien, Kagerer, Contreras-Vidal, & Clark, 2008; Chang & Yu, 2010; Rosenblum & Livneh-Zirinski, 2008; Smits-Engelsman Niemeijer, & van Galen, 2001); Dysgraphia (e.g., Kushki, Schwellnus, Ilyas, & Chau, 2011; Overvelde & Hulstijn, in press; Rosenblum, Dvorkin, & Weiss, 2006; Smits-Engelsman & van Galen, 1997); Huntington’s Disease (e.g., Phillips et al., 1996; Phillips, Chiu, Bradshaw, & Iansek, 1995; Slavin et al., 1999; Yaguez, Canavan, Lange, & Homberg, 1999); Learning Disability (e.g., Galli et al., 2011; van Roon, Caeyenberghs, Swinnen, & Smits-Engelsman, 2010); Schizophrenia (e.g., Grootens et al., 2009; Jahn et al., 2006; Putzhammer et al., 2005; Tigges et al., 2000); and Parkinson’s Disease (e.g., Gangadhar et al., 2009; Poluha, Teulings, & Brookshire, 1998; Ponsen et al., 2006; Rand, Stelmach, & Bloedel, 2000; van Gemmert, Teulings, & Stelmach, 1998). Germane to the present study are findings related to ADHD.
Using qualitative variables such as legibility; spacing, letter size, and alignment consistency; organization of material within space; and letter insertions, transpositions, substitutions, and omissions, studies of handwriting produced by children diagnosed with ADHD indicate that their writing quality is generally poor, immature, and error-prone when compared with non-ADHD controls (Adi-Japha et al., 2007; Flapper et al., 2006; Lerer, Artner, & Lerer, 1979; Marcotte & Stern, 1997; Whalen, Henker, & Finck, 1981; Tucha & Lange, 2001). In addition, poor qualitative performance does not appear to be the result of purely linguistic, visual, perceptual, or visual motor integration deficits (Adi-Japha et al., 2007; Marcotte & Stern, 1997) and typically improves after taking prescribed dosages of stimulant medication (Lerer et al., 1979; Tucha & Lange, 2001; Whalen et al., 1981). Interestingly, kinematic analyses assessing objective, process related aspects of handwriting indicate that the handwriting produced by children diagnosed with ADHD is more dysfluent and thus appears less automatized when taking stimulant medication compared to when they are not taking prescribed medication, and is more dysfluent when such children are on stimulant medication than observed in controls (Flapper et al., 2006; Tucha & Lange, 2001, 2004, 2005). This pattern of fluency and dysfluency related to medication status, however, has not been observed in adults diagnosed with ADHD under similar conditions (Tucha & Lange, 2004). In these contexts, writing fluency is operationalized as the number of changes in direction of velocity or acceleration as recorded by digitizing technology and analyzed by appropriate software. Velocity profiles of fluent, automatized handwriting appear as smooth asymmetrical bell-shaped curves with few changes in velocity/acceleration direction, whereas dysfluent, unautomatized handwriting evinces velocity profiles with multiple “jagged peaks” and
many changes in the direction of velocity/acceleration. See Figures 1 and 2 for examples of fluent versus dysfluent vertical velocity profiles, respectively.

Figure 1. Velocity profile of the word “hello” written fluently.
In addition, these studies demonstrated that while off prescribed dosages of stimulant medication, kinematic measures of graphomotor fluency in affected children were not significantly different from those of non-ADHD controls. Further, it does not appear that these findings are due to a direct effect of medication, as fluent movements can be elicited from children with ADHD taking stimulant medication (Tucha & Lange, 2004). Rather, this decreased fluency and automaticity may be the result of a secondary effect resulting from enhanced attention, from greater cognitive control (Tucha & Lange, 2004; Tucha, Mecklinger, Walitza, & Lange, 2006; Tucha et al., 2003), or from possibly other cognitive, motor, or psychomotor processes influenced by stimulant medication.

Alternatively, Lange et al. (2007) suggested that children and adults with ADHD may, in general, “have difficulties in skills whose acquisition starts as a [laboured] and conscious learning process that becomes automatic following consistent and frequent practice” (p.
Similarly, Flapper et al. (2006) noted that typically, accuracy is achieved before speed and fluency when learning a complex task. In turn, children with ADHD would first need to engage sufficient attentional resources and motor skills for an extended period of time before generating handwriting that is both fluent and accurate, noting that both attentional abilities and motor skills are reported to improve with methylphenidate treatment in adults, adolescents, and children diagnosed with ADHD (Bart, Podoly, & Bar-Haim, 2010; Lerer et al., 1979; Shafritz, Marchione, Gore, Shaywitz, & Shaywitz, 2004; Stray, Stray, Iverson, Ruud, & Ellersten, 2009; Tucha, Mecklinger, Laufkotter, et al., 2006; Tucha, Prell, et al., 2006).

The Present Study

There are few studies that have investigated the kinematic aspects of writing in adults diagnosed with ADHD, with no study specifically examining the potential variability of task performance within the kinematic aspects of graphomotor skills in adults diagnosed with ADHD, and no study comparing novel versus putatively automatized graphomotor processes in this population. As such, using a digitizing tablet to capture kinematic aspects of handwriting, the present study seeks to determine within the context of medication status 1) whether the variability of performance observed in other psychological domains (e.g., task persistence, emotion, and attention) in those diagnosed with ADHD manifests within kinematic variables associated with consistency, stability, and coordination during the execution of an automatized graphomotor task; and 2) assess the effects of novelty on consistency measures of graphomotor performance between adults with and without ADHD. Under the premise that handwriting output is generated from a velocity control perspective (i.e., that the central nervous system
produces volitional graphomotor output by controlling the velocity of an end-effector via interactions between higher-order cortical and sub-cortical systems and lower-level agonist and antagonist neuromuscular systems [Guerfali & Plamondon, 1997; Plamondon, 1993, 1995a, 1995b, 1998; Plamondon, Feng, & Woch, 2003]) and noting variability of performance/behaviour demonstrated in other psychological domains in those diagnosed with ADHD, it is hypothesized that 1) greater intra-individual variability in kinematic velocity measures will be observed in adults diagnosed with ADHD off medication when compared to neurotypical adults. Additionally, 2) although no a priori hypothesis is salient with regard to the effects of novelty on variability measures in those diagnosed with ADHD, it could be speculated that if variability of performance observed in adults with ADHD extends to the graphomotor domain, ADHD participants discontinuing medication will be differentially affected by a novel graphomotor task and in turn elicit greater levels of inconsistency compared to those without ADHD. Should statistically and practically significant differences become evident (i.e., differences of medium to large effect sizes), this would be the first study utilizing kinematic analysis to explicitly demonstrate variability of performance within the graphomotor domain in adults diagnosed with ADHD. Significant results indicating variability in kinematic performance would also add to the current literature indicating that ADHD is not simply a disorder of childhood, but rather, a disorder in which specific motor control differences extend into adulthood. Further, the results of this study would support conducting future research into the use of digitizing technology as an objective diagnostic and descriptive tool within the ADHD population, which in turn may enhance the specificity and/or sensitivity of current assessment and diagnostic techniques.
CHAPTER III
DESIGN AND METHODOLOGY

Participants

Power analysis ($\alpha = .05$, $[1 - \beta] = .80$) indicated that using the proposed methodological design and statistical analysis, 52 total participants would be needed to detect a statistically significant difference of large effect size. For within-group differences, power analysis indicated that 16 participants would be needed to detect differences of large effect size.

Thirty-eight participants were recruited through three sources: control participants were recruited via the University of Windsor’s Psychology Participant Pool ($n = 31$) and clinical participants were recruited through Student Disabilities Services (Education Development Center) at the University of Windsor and through the private practice of a local psychiatrist ($n = 8$). One control participant, however, requested that their data be removed from the study, resulting in a net of 30 control participants and 38 total participants. To minimize confounds related to extraneous visual and motor disturbances, participants included only those with normal or corrected to normal vision and those who did not have an existing neurological condition that would negatively affect graphomotor performance (e.g., cerebral palsy affecting the upper extremities, severe tendinitis, or carpal tunnel syndrome). In addition, clinical participants included only those who were currently taking prescribed dosages of stimulant medication for the treatment of ADHD symptoms. Participants recruited through the University of Windsor’s Psychology Participant Pool received course bonus points (1 point for control participants based on one hour of participation time and 2 bonus points for clinical
Participants based on two hours of participation time) for participating in the research study. Participants recruited through Student Disability Services at the University of Windsor and the private practice of the local psychiatrist received a $10 gift card and a chance to win one of two $50 debit cards via entry into a draw.

**Materials and Apparatus**

**Demographic information and ADHD symptomatology.** For the purposes of sample description, participant demographic information including age, sex, handedness, current medications (including type and dosage), ethnicity, official ADHD diagnosis and subtype (if applicable), and neurological status was collected from each participant via an in-person interview (see Appendix A for the interview form used). For participants diagnosed with ADHD, records pertaining to official diagnoses were reviewed and specific diagnoses if available (e.g., ADHD-C, ADHD-PI, ADHD-HI, and any comorbid diagnoses) were also recorded for descriptive purposes. In addition, all participants completed the Barkley Adult ADHD Rating Scale-IV (BAARS-IV; Barkley, 2011). Based on DSM-IV diagnostic criteria, the BAARS-IV is a self-report questionnaire designed to evaluate current and/or childhood ADHD symptoms in adults between the ages of 18 and 81 years. According to the manual, the normative sample used to develop the BAARS-IV, which consisted of 1,249 adults between the ages of 18 and 96, “closely approximated the U.S. adult population based on the U.S. Census from the year 2000 concerning regional distribution, sex, race/ethnic group, marital status, employment status, total household income, and education” (Barkley, 2011, p. 14). After completion of the questionnaire, a total ADHD score, symptom count, and subscale scores for both current symptoms and childhood symptoms can be calculated. (Note: the BAARS-IV
also contains forms allowing current and childhood symptomatology scores to be derived based on reports from others through the use of an alternative quick-screen. These were not utilized in this study). The BAARS-IV also produces subscale scores related to four recognized ADHD symptom dimensions: Inattention, Hyperactivity, Impulsivity, and Sluggish Cognitive Tempo (SCT). According to the BAARS-IV manual, ADHD scores at or above the 93rd percentile may be interpreted as reflecting a significant abnormality and clinical significance in that domain. Because clinical participants participated in the research both on and off of their ADHD medication, they were asked to answer the questionnaire regarding their current symptomatology within the context of being off of their medication. Finally, if subtype identifier information was unavailable or unknown, a determination of subtype was made based upon the clinical participant’s self-report current ADHD symptoms as measured by the BAARS-IV. That is, for clinical participants only, a subtype identifier of ADHD-PI was given if significant abnormality was reported only within the Inattention domain, a subtype identifier of ADHD-HI was given if significant abnormality was reported only within the Hyperactivity or Impulsivity domains, and a subtype identifier of ADHD-C was given if significant abnormality was reported within both the Inattention domain and the Hyperactivity or Impulsivity domains.

Internal consistency reliability of the BAARS-IV was reported by the manual to be “satisfactory” for current symptom total score and for each subscale/domain score for both current and childhood reported symptoms. Test-retest reliability was described as “reasonable” over a 2- to 3-week period. Finally, construct validity, discriminant validity, and criterion validity are reported to be “satisfactory.” Internal consistency and
test-retest reliability measures for the BAARS-IV, as indicated by the manual (Barkley, 2011), are listed in Table 1.

Table 1

*Internal Consistency & Test-Retest Reliabilities of the BAARS-IV*

<table>
<thead>
<tr>
<th></th>
<th>Internal Consistency Reliability</th>
<th>Test-Retest Reliability**</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Current Symptoms</td>
<td>Childhood Symptoms</td>
</tr>
<tr>
<td>ADHD Inattention</td>
<td>.902</td>
<td>.940</td>
</tr>
<tr>
<td>ADHD Hyperactivity</td>
<td>.776</td>
<td>.912†</td>
</tr>
<tr>
<td>ADHD Impulsivity</td>
<td>.807</td>
<td></td>
</tr>
<tr>
<td>Total Score</td>
<td>.914</td>
<td>.947</td>
</tr>
</tbody>
</table>

*Note.* *Barkley Adult ADHD Rating Scale-IV.* **Test-retest reliability over a 2- to 3-week period. † Represents combined dimension of hyperactivity-impulsivity.

**Estimate of intellectual ability.** An estimate of IQ was derived using four subtests of the Wechsler Adult Intelligence Scale-Fourth Edition (WAIS-IV; Wechsler, 2008). Based on practical considerations and the best combination of short form reliability and validity coefficients (.953 and .940, respectively; see Sattler & Ryan, 2009), the four subtests used for estimating IQ were Block Design (BD), Vocabulary (VC), Arithmetic (AR), and Coding (CD). IQ estimates were used for descriptive purposes and for identifying initial group differences between the control group and the clinical group.

**Kinematic analysis and digitizing tablet.** A WACOM Cintiq 21UX digitizing tablet was used to record the handwriting movements of participants. The digitizing tablet has an active display area of 17” by 12.75” and spatial resolution of 5080 lines per inch. Because this tablet provides real-time on-screen visual feedback, a special non-
inking pen was used by participants. MovAlyzeR software (NeuroScript, LLC; Tempe, AZ, USA) was utilized to quantify handwriting movements with a maximum sampling rate of 200 Hz and x-y coordinates were low-pass filtered at 12 Hz. Handwriting movements were broken down by MovAlyzeR software into strokes using interpolated vertical velocity zero crossings. In this sense, a stroke, representing a “unit” of handwriting, can be defined as “a segment bounded by time moments at which the vertical component of the velocity changes sign” (Teulings, Thomassen, & van Galen, 1983, p. 168).

Kinematic variables derived using MovAlyzeR software include Relative intra-individual standard deviation of Peak Velocity (RPV) and Normalized Jerk (NJ). The RPV variable is a coefficient of variation (CV) that was derived by dividing the absolute standard deviation of mean peak velocity of each digitized word or symbol by the average peak velocity of the digitized word or symbol (Mergl et al., 1999). The word “hello” and the novel symbol “ể” are described below (see Figure B1 in Appendix B for a scaled version of the novel symbol). The RPV variable reflects stability, coordination, and consistency of an individual’s handwriting, with less consistently controlled movements indicated by higher values and more consistently controlled movements reflected by lower values (Mergl et al., 1999; Schroter et al., 2003). An RPV value of 0 would indicate completely identical mean peak velocity across all trials of a writing task.

NJ is a measure of writing smoothness and fluency. High NJ scores indicate dysfluent movement and low NJ scores indicate smoother, fluent, and more automatized movement (Teulings et al., 1997; Yan et al., 2008). Said another way, as one practices
and in turn automatizes a graphomotor program, dysfluency decreases (Portier & van Galen, 1992), as will the NJ variable. In turn, the NJ measure should indicate greater dysfluency when individuals write a novel symbol or grapheme on the digitizing tablet versus a well-practiced and automatized symbol or grapheme. The NJ variable is similar to the dysfluency measure of “number of inversions of acceleration” used in much of the research utilizing kinematics to investigate graphomotor problems in those diagnosed with ADHD (for examples, see Flapper et al., 2006; Schoemaker et al., 2005; Tucha & Lange, 2001, 2004, 2005; Tucha et al., 2006; and Tucha et al., 2003) in that NJ “is the change of acceleration per time” (Teulings et al., 1997, p. 160). NJ, however, has the advantage of allowing the comparison of words or symbols of varying size and movement durations because it is normalized (Teulings et al., 1997).

All demographic and research data were kept confidential and secure. Additionally, participant demographic and research data were de-identified (i.e., coded with a randomly assigned participant identification number) but still attached to identifying information for two weeks after the data were collected, thus giving participants the opportunity to withdraw their data from the study. After this time, the link connecting identifying personal information with demographic and research data was removed and only arbitrary participant identification numbers were associated with demographic and research data.

**Procedures**

In the following order, participants: 1) took part in an interview with the researcher to provide demographic and medical information, 2) answered questions related to ADHD symptomatology, 3) participated in an abbreviated test of general
intellectual ability, 4) signed their name on the digitizing tablet 10 times, 5) wrote the word “hello” in lower-case using cursive handwriting on the digitizing tablet 30 times (representing the automatized condition), and 6) wrote the novel symbol “” on the digitizing tablet 30 times (representing the novel word condition). A sample of this word and symbol was visible to the participant on a card throughout the graphomotor task. Instructions for all tasks were given aurally, with instruction provided visually on the digitizing tablet throughout.

All data from control participants was collected in one session. Data obtained from clinical participants was collected on two occasions, once while the participants were taking prescribed dosages of ADHD medication and a second time after abstaining from prescribed dosages of ADHD medication for a 24 to 48 hour period (withdrawal of medication time-frame based on product information indicating extremely low mean drug plasma concentrations between 24 and 48 hours after taking stimulant medication; U.S. Food and Drug Administration, 2007). The time-frame between test and retest for this group was approximately one week ( = 6.75, = 0.71). The demographics questionnaire, BAARS-IV, and WAIS-IV subtests were completed while the clinical participants were taking prescribed dosages of ADHD medication to minimize potential discomfort associated with the return of ADHD symptomatology combined with a relatively long research process. The Current Symptoms form of the BAARS-IV questionnaire was completed while clinical participants were off of their prescribed ADHD medication. Experimental task administration within the context of medication status was counterbalanced so that half of the clinical participants completed the writing
tasks while taking prescribed medication on their first visit, while the other half completed the writing tasks while taking prescribed medication during their second visit.

To become familiarized with using the digitizing tablet and pen, all participants began the writing task by signing their name on the digitizing tablet 10 times. Subsequently, participants began writing experimental trials. No specific instructions were given related to the quality of the handwriting participants were to produce beyond pointing to the sample and telling the participants to “Write the word hello in cursive and lower case as it is written on the card. Just write how you typically write.” When the researcher was pressed further for additional instruction, participants were only told to “Simply write how you typically write in cursive.” In the case of the novel symbol, participants were instructed as follows: “Here is another symbol for the word ‘hello.’ Please write the symbol as demonstrated on the card.” If participants questioned whether neatness was required, the investigator stated, “Just write it how you would write any other word, but make it look like the symbol as demonstrated on the card.” Because handwriting is variable within individuals, even when writing the same grapheme, participants wrote each word and symbol 30 times in order to acquire a statistically stable sample of handwriting. Finally, all participants were given the ability to manipulate the position of the tablet to one that was comfortable for writing, as well as position the cards containing the word “hello” and the novel symbol wherever was best for them.
CHAPTER IV

ANALYSIS OF RESULTS

All data analyses were performed using IBM SPSS Statistics, Version 20.0. Unless otherwise noted, an alpha level of .05 was used to determine statistical significance. In addition, interpretations of effect sizes using $\omega^2$ were based on Kirk’s (2003) guidelines, such that 0.010 was interpreted as a small association, 0.059 as a medium association, and 0.138 or larger as a large association.

Data Analysis of Assumptions

Prior to hypothesis testing, the data were analyzed to determine adherence to the assumptions of ANOVA, repeated measures ANOVA, and mixed design ANOVA. Cumulatively, tested assumptions included normality of distribution and homogeneity of variance. The assumptions of sphericity and homogeneity of variance/covariance matrices by group were not analyzed due to the research design only incorporating two levels of repeated measures. Assumptions were analyzed using the variables Estimated IQ, Current Total ADHD Score, Childhood Total ADHD Score, NJ under the automatized writing task condition, and RPV under both the automatized and novel writing task conditions, with group membership (i.e., control versus clinical participants) as the independent variable (IV).

Homogeneity of variance. Homogeneity of variance was assessed by first identifying outlier variables (i.e., data with derived $z$-scores greater than |2.5|) and next using Levene’s test of equality of error variances, with statistical significance of the latter (i.e., $p < .05$) reflecting a potential violation of this assumption. The following outliers were identified: one control participant within the Estimated IQ dependent variable (DV)
and one clinical participant within both the NJ DV during automatized graphomotor execution while on ADHD medication as well as within the NJ DV during automatized graphomotor execution while off ADHD medication. In turn, all subsequent analyses of assumptions were conducted with and without the inclusion of outliers for comparative purposes. A significant Levene’s statistic was found between the variances of the ADHD group on medication and control participants within the NJ DV during the automatized writing task. No other statistical significance was found using Levene’s test, indicating that the variation within conditions was roughly equivalent for all other comparisons. When outliers were removed from the dataset, homogeneity of variance statistics improved for the NJ DV during the automatized writing task when comparing control participants versus clinical participants on ADHD medication, but statistical significance persisted. Removing the outlier found within the Estimated IQ DV did not affect the non-significant finding of Levene’s test of homogeneity of variance, and in fact moved the data closer towards heterogeneity of variance.

It is important to note that ANOVA may be robust to violations of homogeneity of variance when comparison groups are equal or nearly equal in size (i.e., the larger group contains less than 1.50 times the number of participants than the smaller group) and when the variance distribution between the largest and smallest variances is not greater than a 4:1 ratio. The control group was 3.75 times larger than the clinical group and 4.29 times larger than the clinical group for comparisons in which outliers were removed. In addition, when outliers were retained, the variance distribution between the largest and smallest variance in NJ data during the automatized writing task comparing the ADHD group on medication and control participants was over 20:1. When outlier
data was removed, the variance ratio in this comparison decreased substantially to almost 5:1. Variance data are presented in Table 2 and the results of homogeneity of variance testing are summarized in Table 3.

Table 2

Variance of Data Within Conditions

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>Group</th>
<th>Variance (With Outliers)</th>
<th>Variance (Without Outliers)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Estimated IQ</td>
<td>Control</td>
<td>122.01</td>
<td>88.92†</td>
</tr>
<tr>
<td></td>
<td>ADHD</td>
<td>201.84</td>
<td>201.84</td>
</tr>
<tr>
<td>Current ADHD</td>
<td>Control</td>
<td>63.10</td>
<td>63.10</td>
</tr>
<tr>
<td></td>
<td>ADHD</td>
<td>117.64</td>
<td>117.64</td>
</tr>
<tr>
<td>Childhood ADHD</td>
<td>Control</td>
<td>60.44</td>
<td>60.44</td>
</tr>
<tr>
<td></td>
<td>ADHD</td>
<td>54.12</td>
<td>54.12</td>
</tr>
<tr>
<td>NJ – Auto</td>
<td>Control</td>
<td>50.66</td>
<td>50.66</td>
</tr>
<tr>
<td></td>
<td>On Rx</td>
<td>1028.50</td>
<td>243.96†</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>715.81</td>
<td>98.98†</td>
</tr>
<tr>
<td>RPV – Auto</td>
<td>Control</td>
<td>0.0015</td>
<td>0.0015</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>0.0019</td>
<td>0.0019</td>
</tr>
<tr>
<td>RPV – Novel</td>
<td>Control</td>
<td>0.0026</td>
<td>0.0026</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>0.0037</td>
<td>0.0037</td>
</tr>
</tbody>
</table>

Note. IQ = Estimated Full Scale IQ; NJ = Normalized Jerk; RPV = Relative Intraindividual Peak Velocity. Rx = Clinical/ADHD participants’ medication status (On or Off medication). † = a change in value from “with outliers” to “without outliers.”
## Table 3

*Levene’s Test of Homogeneity of Variance*

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>Group Comparison</th>
<th>F</th>
<th>df</th>
<th>Sig.</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>w</td>
<td>w/o</td>
<td>w</td>
</tr>
<tr>
<td>Estimated IQ</td>
<td>Control vs. ADHD</td>
<td>0.58</td>
<td>1.56</td>
<td>1, 36</td>
</tr>
<tr>
<td>Current ADHD</td>
<td>Control vs. ADHD</td>
<td>0.14</td>
<td>0.14</td>
<td>1, 36</td>
</tr>
<tr>
<td>Childhood ADHD</td>
<td>Control vs. ADHD</td>
<td>0.04</td>
<td>0.04</td>
<td>1, 36</td>
</tr>
<tr>
<td>NJ – Auto</td>
<td>Control vs. On Rx</td>
<td>22.50</td>
<td>10.02†</td>
<td>1, 36</td>
</tr>
<tr>
<td></td>
<td>Control vs. Off Rx</td>
<td>12.40</td>
<td>2.75†</td>
<td>1, 36</td>
</tr>
<tr>
<td></td>
<td>On Rx vs. Off Rx</td>
<td>0.30</td>
<td>1.25†</td>
<td>1, 14</td>
</tr>
<tr>
<td>RPV – Auto</td>
<td>Control vs. Off Rx</td>
<td>0.22</td>
<td>0.22</td>
<td>1, 36</td>
</tr>
<tr>
<td>RPV – Novel</td>
<td>Control vs. Off Rx</td>
<td>0.17</td>
<td>0.17</td>
<td>1, 36</td>
</tr>
</tbody>
</table>

*Note.* Analyses of the assumption of homogeneity of variance within the data. IQ = Estimated Full Scale IQ; NJ = Normalized Jerk, RPV = Relative Intraindividual Peak Velocity; Auto = automatized writing condition; Rx = ADHD participants’ medication status (On or Off medication); w = results with outliers; w/o = results without outliers. * = statistical significance (p < .05) and violation of the assumption of homogeneity of variance. † = a change in value from “with outliers” to “without outliers.”

**Normality and Independence of Observations.** The assumption of normality was tested by analyzing skewness, kurtosis, and Shapiro-Wilk statistics. With the inclusion of outlier scores, skewness z-scores were outside conventional cut-offs of significance (i.e., skewness greater than |2|) for the NJ DV within the control group under the automatized condition as well as for the NJ DV within the clinical group off ADHD medication, both of which indicated a positive skew and a potential violation of the assumption of normality. All other cells did not reflect significant positive or negative skewness. Kurtosis z-scores, however, were greater than conventional cutoffs (i.e.,
kurtosis greater than |3|) for the NJ DV under the automatized writing task for controls, clinical participants on ADHD medication, and clinical participants off ADHD medication, indicating significant leptokurtic kurtosis. The Shapiro-Wilk statistic was also statistically significant for these same cells as well as for the Current Total ADHD Score as measured by the BAARS-IV. Within the RPV DV, no significant skewness or kurtosis was observed, and the Shapiro-Wilk statistic was non-significant for all data cells for this variable. This in turn indicated that RPV results represented the only normally distributed experimental data. Removing outlier data resulted in normalizing the distribution of data with regard to skewness and kurtosis for all data cells with the exception of the NJ DV under the automatized writing task within the control group, which retained its significantly positive skew and significantly leptokurtic distribution. In addition, the NJ DV under the automatized writing task continued to produce a significant Shapiro-Wilk statistic within the clinical participant group on ADHD medication.

ANOVA is said to be robust to violations of normality when sample sizes are large and group sizes are roughly equivalent. Noting the relatively small sample size and large group size differences within the sample, the violations of normality found within the NJ DV while retaining outlier variables would significantly impact the reliability of the ANOVA F statistic for all comparisons involving the NJ DV. Removing outlier data normalized the distribution of NJ results for clinical participants off ADHD medication, but not for NJ results for the control group or clinical participants on ADHD medication. Tests of normality data, with and without outliers, are summarized in Table 4.
Table 4

**Normality of Data**

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th></th>
<th></th>
<th></th>
<th>Skewness</th>
<th>Kurtosis</th>
<th>Shapiro-Wilk</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Group</td>
<td>w</td>
<td>w/o</td>
<td>w</td>
<td>w/o</td>
<td>w</td>
</tr>
<tr>
<td>Estimated IQ</td>
<td>Control</td>
<td>0.58</td>
<td>-0.14†</td>
<td>1.03</td>
<td>-0.70†</td>
<td>.34</td>
</tr>
<tr>
<td></td>
<td>Clinical</td>
<td>-0.52</td>
<td>-0.52</td>
<td>-0.12</td>
<td>-0.12</td>
<td>.78</td>
</tr>
<tr>
<td>Current Total ADHD Score</td>
<td>Control</td>
<td>1.10</td>
<td>1.10</td>
<td>1.58</td>
<td>1.58</td>
<td>.03*</td>
</tr>
<tr>
<td></td>
<td>Clinical</td>
<td>0.30</td>
<td>0.30</td>
<td>-1.72</td>
<td>-1.72</td>
<td>.20</td>
</tr>
<tr>
<td>Childhood Total ADHD Score</td>
<td>Control</td>
<td>0.02</td>
<td>0.02</td>
<td>-1.14</td>
<td>-1.14</td>
<td>.23</td>
</tr>
<tr>
<td></td>
<td>Clinical</td>
<td>-0.40</td>
<td>-0.40</td>
<td>-1.14</td>
<td>-1.14</td>
<td>.31</td>
</tr>
<tr>
<td>NJ – Auto</td>
<td>Control</td>
<td>2.28*</td>
<td>2.28*</td>
<td>7.63*</td>
<td>7.63*</td>
<td>.00*</td>
</tr>
<tr>
<td></td>
<td>On Rx</td>
<td>1.89</td>
<td>1.40†</td>
<td>3.53†</td>
<td>0.73†</td>
<td>.01*</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>2.19*</td>
<td>1.11†</td>
<td>5.09*</td>
<td>0.44†</td>
<td>.00*</td>
</tr>
<tr>
<td>RPV - Auto</td>
<td>Control</td>
<td>0.02</td>
<td>0.02</td>
<td>0.10</td>
<td>0.10</td>
<td>.31</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>1.38</td>
<td>1.38</td>
<td>1.88</td>
<td>1.88</td>
<td>.15</td>
</tr>
<tr>
<td>RPV - Novel</td>
<td>Control</td>
<td>0.42</td>
<td>0.42</td>
<td>-0.89</td>
<td>-0.89</td>
<td>.06</td>
</tr>
<tr>
<td></td>
<td>Off Rx</td>
<td>0.93</td>
<td>0.93</td>
<td>-0.86</td>
<td>-0.86</td>
<td>.06</td>
</tr>
</tbody>
</table>

*Note.* Analyses of the assumption of normality of distribution within the data. NJ = Normalized Jerk; RPV = Relative Intraindividual Peak Velocity; Auto = Automatized writing task; Rx = ADHD participants’ medication status (On or Off medication); w = results with outliers; w/o = results without outliers. * = statistical significance (p < .05) and violation of the assumption of normality. † = a change in value from “with outliers” to “without outliers.”

Finally, data were gathered from participants in individual sessions. Combined with the general novelty of the experimental tasks utilized, lack of known organized communication between participants, and the manner in which data were gathered, it is unlikely that participants’ scores were systematically related.

Taken together, the non-normally distributed data on the NJ DV for control participants and clinical participants on ADHD medication combined with the
heterogeneity of variance found comparing these two groups increases the risk of Type 1 errors when conducting analyses based on NJ data. In addition, largely unequal sample sizes and variance distributions greater than 4:1 make conclusions drawn from NJ DV results during the automatized writing task tenuous due the additional violation in the assumption of normality. Again, removing outlier data improved the normality of the data, but did not eliminate non-normality entirely. Given the above stated issues associated with the data and the assumptions of ANOVA, 1) all subsequent analyses were conducted without the presence of outlier scores on the NJ DV and 2) nonparametric statistical analyses were also conducted for comparisons between control and clinical participants on ADHD medication that involve using the NJ DV in order to provide support for, or against, significant findings that were found using parametric statistics.

Demographics and ADHD Symptomatology

Data pertaining to participant demographics, estimated IQ, and presence of ADHD symptomatology were collected for the purposes of sample description. ANOVAs were performed to determine significant initial group differences when appropriate.

In the overall sample, more participants were right-handed (84.21%) than left-handed (15.79%), more women (78.95%) participated in the study than men (21.05%), and a majority of the participants self-identified as Caucasian (71.05%). Between the control and clinical participant groups, a greater proportion of clinical participants were left-handed (37.50%) than control participants (10.00%). Fortunately, kinematic variables have not been shown to be affected by handedness alone (Mergl et al., 1999). There was a greater proportion of men in the clinical participant sample (62.50%) than in
the control participant sample (10.00%). A slightly greater representation of non-dominant ethnic/racial group members was also observed in the clinical participant sample (33.33%) versus the control participant sample (20.00%).

Overall, control participants ($M = 27.56, SD = 11.91$) were younger than clinical participants ($M = 35.00, SD = 9.08$) participants, but not significantly, $F(1, 36) = 2.69, p = .110$, and with a small effect size, $\omega^2 = .043$. There was, however, a broader age range in control participants (age range: 18.58–54.08 years) versus clinical participants (age range: 23.25–46.60 years). Control ($M = 94.17, SD = 11.05$) and ADHD ($M = 94.88, SD = 14.21$) participants performed nearly identically on the general test of intellectual ability as a group, $F(1, 36) = 0.02, p = .880, \omega^2 = .000$, with both groups falling within the average range. Please see Table 5 for a summary of all participant demographic information.
### Participant Demographic Information

<table>
<thead>
<tr>
<th></th>
<th>Control</th>
<th></th>
<th>ADHD</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>n</td>
<td>M</td>
<td>SD</td>
<td>n</td>
</tr>
<tr>
<td>Handedness</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Right</td>
<td>27</td>
<td>-</td>
<td>-</td>
<td>5</td>
</tr>
<tr>
<td>Left</td>
<td>3</td>
<td>-</td>
<td>-</td>
<td>3</td>
</tr>
<tr>
<td>Sex</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Women</td>
<td>27</td>
<td>-</td>
<td>-</td>
<td>3</td>
</tr>
<tr>
<td>Men</td>
<td>3</td>
<td>-</td>
<td>-</td>
<td>5</td>
</tr>
<tr>
<td>Race/Ethnicity</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Asian</td>
<td>2</td>
<td>-</td>
<td>-</td>
<td>0</td>
</tr>
<tr>
<td>Black</td>
<td>3</td>
<td>-</td>
<td>-</td>
<td>1</td>
</tr>
<tr>
<td>Caucasian</td>
<td>24</td>
<td>-</td>
<td>-</td>
<td>6</td>
</tr>
<tr>
<td>Hispanic</td>
<td>1</td>
<td>-</td>
<td>-</td>
<td>1</td>
</tr>
<tr>
<td>Age</td>
<td>-</td>
<td>27.56</td>
<td>11.91</td>
<td>-</td>
</tr>
<tr>
<td>Estimated IQ</td>
<td>-</td>
<td>94.17</td>
<td>11.05</td>
<td>-</td>
</tr>
</tbody>
</table>

**Note.** Estimated IQ = estimate of general intellectual ability. ADHD = clinical participants diagnosed with ADHD.

Participant ratings of ADHD symptomatology as measured by the BAARS-IV are summarized in Table 6. Clinical participants rated current ADHD symptomatology ($M = 48.88$, $SD = 7.86$) as occurring significantly more frequently than control participants ($M = 30.00$, $SD = 7.94$), $F(1, 36) = 35.80, p < .001$, with an observed large effect size, $\omega^2 = .478$. In addition, clinical participants reported significantly more symptoms of ADHD that occurred during childhood ($M = 50.88$, $SD = 7.36$) than did control participants ($M = 30.67$, $SD = 7.77$), $F(1, 36) = 43.56, p < .001$, also with a large effect size, $\omega^2 = .528$. 
Table 6

**BAARS-IV ADHD Symptomatology**

<table>
<thead>
<tr>
<th></th>
<th>Control</th>
<th>Clinical</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>M</td>
<td>SD</td>
</tr>
<tr>
<td>Current Symptom Total</td>
<td>30.00</td>
<td>7.94</td>
</tr>
<tr>
<td>ADHD Score:</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Childhood Symptom</td>
<td>30.67</td>
<td>7.77</td>
</tr>
<tr>
<td>Total ADHD Score:</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

As shown in Table 7, a greater proportion of clinical participants also reported clinically significant levels of current (100%) and childhood (87.50%) ADHD symptomatology compared to control participants (significant levels of current ADHD symptomatology = 13.3%, significant levels of childhood ADHD symptomatology = 0%). Taken together, it can be reasonably concluded that clinical participants reported significantly higher levels of current and childhood ADHD symptoms than control participants. Finally, ADHD subtypes diagnosed in clinical participants included ADHD-C (n = 5), ADHD-PI (n = 1), and ADHD-HI (n = 2).

Table 7

**Number of Participants with BAARS-IV Scores ≥ 93rd %tile**

<table>
<thead>
<tr>
<th></th>
<th>Control</th>
<th>Clinical</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>n</td>
<td>n</td>
</tr>
<tr>
<td>Current Symptom Total</td>
<td>4 (13.30%)</td>
<td>8 (100%)</td>
</tr>
<tr>
<td>ADHD Score:</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Childhood Symptom</td>
<td>0 (0.00%)</td>
<td>7 (87.50%)</td>
</tr>
<tr>
<td>Total ADHD Score:</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Kinematic Analyses

**Overlap with past research.** Due to multiple comparisons and the relatively low risk involved with rejecting a true null hypothesis, a Bonferroni correction was used for all statistical comparisons. As such, the alpha level for statistical comparisons made to determine support for past research was adjusted to .02. Two One-Way ANOVAs were used to compare the handwriting fluency of controls with clinical participants both on ADHD medication as well as off ADHD medication. Between group comparisons based on fluency measures detected no statistically significant differences between control \((M = 18.43, SD = 7.12)\) and clinical participants taking ADHD medication \((M = 25.57, SD = 15.62)\), \(F(1, 35) = 3.45, p = .072\), but did demonstrate a medium effect size, \(\omega^2 = .062\). This non-significant finding was consistent with previous research. Nonparametric statistical analysis (i.e., the Mann-Whitney U Test) also indicated that this difference was not statistically significant, \(p = .435\), and that the null hypothesis should be retained. The handwriting fluency scores of clinical participants off ADHD medication \((M = 21.10, SD = 10.47)\) compared to those of control participants \((M = 18.43, SD = 7.12)\) were not significantly different, \(F(1, 35) = 0.66, p = .421\), and demonstrated an uninterpretable effect size, \(\omega^2 = .000\). Combined, these non-significant findings are consistent with previous research.

A repeated measures One-Way ANOVA was used to examine automatized handwriting fluency (i.e., NJ) of clinical participants taking ADHD medication versus those same participants discontinuing their ADHD medication for 24 hours. The results comparing the handwriting fluency of clinical participants on ADHD medication \((M = 25.57, SD = 15.62)\) versus off ADHD medication \((M = 21.10, SD = 10.47)\) did not
identify a statistically significant difference, $F(1, 6) = 1.51, p = .265$. This result is consistent with previous findings in that handwriting fluency of clinical participants did not differ significantly based on medication status. Although not a statistically significant difference, a medium effect size was found, $\omega^2_{\text{partial}} = .068$. See Table 8 for a summary of automatized fluency results as well as source data for statistical comparisons and Figure 3 for a graphical comparison of the automatized writing fluency of control participants versus clinical participants both on and off ADHD medication.

Table 8

Automatized Writing Fluency – Summary and Source Table

<table>
<thead>
<tr>
<th>Source</th>
<th>$M$</th>
<th>$SD$</th>
<th>$SS$</th>
<th>$df$</th>
<th>$MS$</th>
<th>$F$</th>
<th>$p$</th>
<th>$\omega^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Group Membership</td>
<td>-</td>
<td>-</td>
<td>289.14</td>
<td>1</td>
<td>289.14</td>
<td>3.45</td>
<td>.072</td>
<td>.062</td>
</tr>
<tr>
<td>Error Between</td>
<td>-</td>
<td>-</td>
<td>2932.99</td>
<td>35</td>
<td>83.80</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Control</td>
<td>18.43</td>
<td>7.12</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>On Rx</td>
<td>25.57</td>
<td>15.62</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Group Membership</td>
<td>-</td>
<td>-</td>
<td>40.24</td>
<td>1</td>
<td>40.24</td>
<td>0.66</td>
<td>.421</td>
<td>.000</td>
</tr>
<tr>
<td>Error Between</td>
<td>-</td>
<td>-</td>
<td>2127.38</td>
<td>35</td>
<td>60.78</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Control</td>
<td>18.43</td>
<td>7.12</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Off Rx</td>
<td>21.01</td>
<td>10.47</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Medication Status</td>
<td>-</td>
<td>-</td>
<td>70.09</td>
<td>1</td>
<td>70.09</td>
<td>1.51</td>
<td>.265</td>
<td>.068†</td>
</tr>
<tr>
<td>Error Residual</td>
<td>-</td>
<td>-</td>
<td>278.44</td>
<td>6</td>
<td>46.41</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>On Rx</td>
<td>25.57</td>
<td>15.62</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Off Rx</td>
<td>21.10</td>
<td>10.47</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

*Note.* On Rx = ADHD/Clinical participants on ADHD medication; Off Rx = ADHD/Clinical participants off ADHD medication. $M$ = mean of normalized jerk (NJ) value; $SD$ = standard deviation of normalized jerk (NJ) value. † = partial omega-squared.
Variability of graphomotor performance. The alpha level was set at .05 to indicate statistical significance for the following comparisons. A One-Way ANOVA was used to analyze the kinematic variability of automatized handwriting performance (i.e., RPV) of healthy control participants with that of ADHD participants that discontinued ADHD medication for 24 hours. No main effect was found related to group membership and variability of graphomotor performance, $F(1, 36) = 0.37$, $p = .545$, and no interpretable effect size was found, $\omega^2 = .000$, indicating that clinical participants not taking ADHD medication ($M = 0.12$, $SD = 0.06$) demonstrated similar variability in automatized graphomotor performance to control participants ($M = 0.13$, $SD = 0.04$). See Table 9 for a summary of graphomotor variability findings as well as source information. See Figure 4 for a graphical comparison of graphomotor variability findings.

![Automatized Writing Fluency](image-url)
Table 9

Automatized Writing Variability – Summary and Source Table

<table>
<thead>
<tr>
<th>Source</th>
<th>M</th>
<th>SD</th>
<th>SS</th>
<th>df</th>
<th>MS</th>
<th>F</th>
<th>p</th>
<th>$\omega^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Group Membership</td>
<td>-</td>
<td>-</td>
<td>.001</td>
<td>1</td>
<td>.001</td>
<td>0.37</td>
<td>.545</td>
<td>.000</td>
</tr>
<tr>
<td>Error Between</td>
<td>-</td>
<td>-</td>
<td>.057</td>
<td>36</td>
<td>.002</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Control</td>
<td>.13</td>
<td>.04</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Off Rx</td>
<td>.12</td>
<td>.04</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

Note. Off Rx = ADHD/Clinical participants off ADHD medication. M = mean of the relative intraindividual mean peak velocity (RPV) value; SD = standard deviation of the relative intraindividual mean peak velocity (RPV) value.

Figure 4. Variability of handwriting performance in control and clinical participants. RPV = Relative Intraindividual Mean Peak Velocity. Less consistently controlled movements are indicated by higher values and more consistently controlled movements reflected by lower values. ADHD Off Rx = ADHD/Clinical participants off ADHD medication.

The effects of novelty on variability. A 2 x 2 factorial mixed design ANOVA was used to compare the effects of novelty on variability measures in those diagnosed with ADHD off medication versus healthy controls. There was a significant main effect
for writing task, $F(1, 36) = 24.86$, $p < .001$, indicating that overall, more variable and less consistent handwriting was seen in participants when performing the novel writing task ($M = 0.18$, $SD = 0.06$) versus the automatized writing task ($M = 0.12$, $SD = 0.04$). The effect size for the writing task main effect was large, $\omega^2_{\text{partial}} = .333$. However, no significant interaction effect was observed, $F(1, 36) = 0.11$, $p = .740$, with no interpretable effect size, $\omega^2_{\text{partial}} = .000$. See Table 10 for source information pertaining to graphomotor variability findings as a function of writing task. See Figure 5 for a graphical comparison of graphomotor variability findings related to writing task.

Table 10

*Graphomotor Variability as a Function of Novelty – Source Table*

<table>
<thead>
<tr>
<th>Source</th>
<th>SS</th>
<th>df</th>
<th>MS</th>
<th>F</th>
<th>p</th>
<th>$\omega^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Within Subjects</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Writing Task</td>
<td>0.040003</td>
<td>1</td>
<td>0.040003</td>
<td>24.86</td>
<td>&lt;.001*</td>
<td>.333</td>
</tr>
<tr>
<td>Error Within</td>
<td>0.057937</td>
<td>32</td>
<td>0.001609</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Interaction</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Writing Task x Group Membership</td>
<td>0.000180</td>
<td>1</td>
<td>0.000180</td>
<td>0.11</td>
<td>.740</td>
<td>.000</td>
</tr>
<tr>
<td>Error Interaction</td>
<td>0.057937</td>
<td>32</td>
<td>0.001636</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

Note.  * = statistical significance $p < .05$
Figure 5. Handwriting variability based on writing task. RPV = Relative Intraindividual Mean Peak Velocity. Less consistently controlled movements are indicated by higher values and more consistently controlled movements reflected by lower values. ADHD Off Rx = ADHD/Clinical participants off ADHD medication.
Discussion

Utilizing a digitizing tablet and specialized kinematic analyses software to quantify graphomotor performance during putatively automatized and novel writing tasks in adults with and without a diagnosis of ADHD, the present study sought to: 1) determine whether the variability of performance observed in other psychological domains in those diagnosed with ADHD would manifest within kinematic variables associated with stability and coordination of graphomotor output (i.e., RPV) and 2) determine whether a novel writing task would differentially affect the variability of graphomotor output of adults diagnosed with ADHD compared to healthy controls.

Overlap with past research. Consistent with prior research (Tucha & Lange, 2004), results of the present study suggest that automatized graphomotor fluency, as measured by kinematic analysis, is not significantly different in adults diagnosed with ADHD taking prescribed dosages of stimulant medication from that of neurotypical adults. This conclusion was supported using both parametric and nonparametric statistical analysis. Despite these non-statistically significant findings that appear to corroborate past research, this conclusion should be accepted cautiously. Findings of statistical significance using ANOVA or other analyses under the general linear model are affected by sample size. That is, as sample size increases, the likelihood of finding a statistically significant result continues to increase, even when differences in performance are relatively small. As such, the power of the research design, which takes into account sample size, effect size, and alpha level, must also be considered. Noting the medium
effect size ($\omega^2 = .062$) and the relatively small number of clinical participants ($n = 7$) within this study, it is likely that this non-statistically significant finding is due to low power rather than non-significant differences between healthy adults and adults diagnosed with ADHD on stimulant medication. In turn, a statistically significant result may have been found if the clinical sample was larger and fluency results within the study maintained the same pattern.

Also consistent with previous research, adults diagnosed with ADHD off medication produced similarly fluent automatized graphomotor output as control participants without ADHD. Although an argument could be made that increased power via a larger sample of clinical participants may result in the formulation of a different conclusion, the extremely small (and uninterpretable) effect size observed in this study combined with both the adherence to the assumptions of ANOVA and findings that are consistent with previous research strongly suggest that this conclusion is reliable and valid.

Concerning the handwriting fluency of adults diagnosed with ADHD on stimulant medication versus those same adults off ADHD medication, the present study was consistent with previous research indicating no statistically significant difference in automatized graphomotor fluency. This finding suggests that medication status may not affect the graphomotor fluency of adults diagnosed with ADHD when performing an automatized writing task. Again, this interpretation must be made with caution. Similar to the above results comparing the automatized graphomotor fluency of control participants with that of clinical participants on ADHD medication, the medium effect size ($\omega^2 = .068$) found in the comparison of adults with ADHD on medication versus off
medication suggests that non-significant findings could be the result of insufficient statistical power associated with a small clinical participant sample.

However, if statistical power were sufficient and statistically significant findings related to differences in automatized graphomotor fluency were found between adults with ADHD on stimulant medication versus off ADHD medication, as well as between adults with ADHD on stimulant medication versus control participants, results would still need to be interpreted cautiously due to other methodological considerations. First, graphomotor fluency in previous research was operationalized as the mean number of inversions in the direction of vertical velocity over time (Tucha & Lange, 2004). In this study, however, graphomotor fluency was operationalized as normalized jerk, which is derived from the number of changes in acceleration in time (analogous to the number of changes in velocity over time) but was then normalized due to the effects of size and duration of movements on fluency measures (Tuelings et al., 1997). Using the NJ variable as opposed to the mean number of inversions in velocity provided the benefit of validly comparing graphomotor fluency of the word “hello” and the symbol “\u010ce\u0103\u0161.” However, the NJ variable may not be completely analogous to previously used measures quantifying automatized graphomotor fluency in adults with ADHD. As such, the derived fluency measures of this study may not be completely comparable with those of past research investigating graphomotor fluency in adults with ADHD.

Secondly, the previous study investigating graphomotor fluency in adults diagnosed with ADHD analyzed the specific letter combination of “ll” within two German words (Tucha & Lange, 2004). The present study, however, analyzed the entire word “hello” to determine automatized graphomotor fluency. Intuitively, the
biomechanical forces necessary to generate the letters “ll” versus the word “hello” are different based upon salient graphemic differences. In turn, conclusions drawn from comparisons between the current study and past research should be done with caution noting that differences in graphomotor fluency findings may be due, at least in part, to the experimental stimulus used.

**Variability of graphomotor performance.** Using the kinematic variable RPV, clinical participants exhibited variability in graphomotor output when executing an automatized writing task that was similar to that of control participants. As such, the results of this study suggest that the variability of performance observed in various psychological domains (e.g., emotional expression, handwriting production, fine motor skill movements, motor coordination, and motor force output) within the ADHD population may not be manifest within the kinematic measures of stability, consistency, and coordination used in this study, in which participants performed an automatized graphomotor task.

This was the first such research to utilize RPV as a measure of variability of graphomotor performance in adults diagnosed with ADHD. Previous research utilizing the RPV measure to study kinematic aspects of handwriting in clinical populations focused on patients diagnosed with probable Alzheimer’s dementia (AD) (Schroter et al., 2003). In this research, Schroter and colleagues found that when performing a spiral drawing task, participants with probable AD exhibited significantly more variability, incoordination, and greater inconsistency in the kinematics of handwriting movements compared to similarly aged healthy control participants. One of the primary rationales for conducting this study was the high co-occurrence of salient motor dysfunction found
within the AD population: a high prevalence of Parkinsonism and various extrapyramidal motor symptoms. Although motor problems have also been documented in those with ADHD, motor sequelae in the ADHD population may be more subtle than what is observed in patients with AD. Accordingly, a failure to demonstrate significant differences in consistency of graphomotor output between adults with ADHD and healthy controls could be the result of less than optimal sensitivity of the RPV measure in detecting subtle motor differences in the ADHD population.

Further, how the RPV measure was derived may have created an insensitive measure that failed to detect intraindividual variability of graphomotor production. That is, the mean peak velocity was collapsed across all strokes within each trial and then averaged with all 30 trials of each writing task. The RPV, which is a coefficient of variation (CV), was then determined for each participant by dividing the standard deviation of the mean peak velocities by the average of the mean peak velocities across all trials. As such, this collapsed mean may not have optimally reflected variability of graphomotor performance with sufficient sensitivity because variability within each individual writing trial was not taken into consideration when calculating the RPV. Said another way, deriving the RPV variable to indicate intraindividual variability by creating a CV based on the averages and standard deviation of mean peak velocities across all trials rather than deriving the RPV variable based upon an average of the coefficients of variation of mean peak velocities calculated within each trial may have underestimated the intraindividual variability of automatized and novel graphomotor output.

The writing tasks themselves may also not have optimally allowed participants to demonstrate significant variability of graphomotor performance. For example, previous
work utilizing the RPV variable has demonstrated no statistically significant differences in intraindividual variability of handwriting between men and women when writing letters or simple geometric figures (Mergl et al., 1999). However, in this same study, statistically significant gender differences in intraindividual variability of graphomotor performance emerged when participants executed a six word sentence.

Bearing in mind the methodological and statistical considerations mentioned above, the findings of the present study suggest the following: 1a) conclusions concerning the fluency of automatized graphomotor performance in healthy controls as compared with adults diagnosed with ADHD taking stimulant medication can only be tentatively drawn at this time. These findings, in combination with past findings, suggest that the automatized graphomotor fluency between these groups is similar. However, medium effect sizes suggest that real differences may be present and would be detected if statistical power were greater. 1b) A negligible observed effect size, non-significant findings, and replication of past research provide strong evidence that the graphomotor fluency of adults with ADHD off stimulant medication is similar to the fluency of individuals without ADHD. 2) Conclusions regarding the fluency of automatized graphomotor performance in adults with ADHD on stimulant medication versus off ADHD medication can only be tentatively drawn. The medium effect size found within this comparison suggests potential differences in automatized graphomotor fluency in adults with ADHD on stimulant medication versus off medication. 3) Variability of performance, as measured by the RPV variable, does not appear to manifest within the graphomotor domain in adults diagnosed with ADHD. However, more sensitive measures of intraindividual variability of graphomotor performance or different writing
tasks may yield different results and in turn, support an alternative conclusion. 4) The novelty of a graphomotor task does not appear to differentially affect the variability of kinematic handwriting performance in adults diagnosed with ADHD as measured by the RPV variable. As mentioned previously concerning the RPV variable, however, different results may be found by calculating a more sensitive measure of intraindividual variability of graphomotor performance.

An additional aim of this study was to add to the current literature demonstrating that ADHD is not simply a disorder of childhood, but rather, a condition that involves specific motor differences that persist into adulthood. Although differences in motor functioning as measured by the kinematic analyses utilized in this study were not statistically significant, effect size differences and methodological considerations do not support the conclusion that motor symptoms do not persist into adulthood. Rather, the results of this study support implementing improved methodology and statistical analyses to further explore potential motor skill differences in those with ADHD that may persist into adulthood.

Further, this study aimed to find support for the use of digitizing technology as an objective diagnostic and descriptive tool within the ADHD population, which would in turn enhance the specificity and/or sensitivity of current assessment and diagnostic techniques. Although significant differences in graphomotor function were not observed between adults with ADHD and healthy controls utilizing the proposed kinematic analyses, when considering the methodological and statistical concerns associated with this study, it is likely too early to conclude that kinematic analysis utilizing digitizing
technology would not create an added benefit when attempting to describe graphomotor performance differences in the adult ADHD population versus unaffected adults.

**Methodological Limitations.** This project began with several pragmatic considerations, including researching and obtaining appropriate hardware and software, developing research protocols, and learning how to interpret the results of kinematic analyses based on digitizing technology. As such, the present study served as a foundation for future kinematic research utilizing digitizing technology in an effort to understand fine motor and graphomotor skill performance of adults and children with ADHD. An additional benefit of this study was to implement a novel protocol and evaluate the feasibility of future research questions. Although this was not a direct goal, the results and implementation of this project did have the benefit of establishing the limitations of kinematic research utilizing digitizing technology. Nevertheless, methodological limitations and limitations of statistical analyses used may have affected the results, and in turn the conclusions of this study.

Beyond the concerns mentioned above, the primary limitation of this study was the small number of clinical participants that were recruited and as such, low statistical power. Post-hoc power analyses using G*Power software (Buchner, Erdfelder, Faul, & Lang, 2009) confirmed the low power of this study, which ranged from $1 - \beta = .19$ to $1 - \beta = .74$ for all statistical analyses. In addition, when sample sizes are small, data are unstable and yield statistics that are unreliable. An additional problem that this small clinical sample size created was a situation in which the two comparison groups were largely unequal in size. This additional problem further decreases the reliability of the $F$ statistic and its ability to help draw conclusions. In the future, recruitment strategies should
involve broadening the pool from which clinical participants are drawn by increasing the number of sources from which participants are recruited. However, the present study could also be considered an initial work in progress, with additional recruitment of participants scheduled to occur in the future.

In addition, parametric statistical analyses in this study utilized the general linear model. These statistical analyses used to determine both group differences in the variability of automatized graphomotor performance and the effect of novelty on these measures of variability may not have been sufficiently sensitive in detecting intraindividual differences. For example, King, Harring, Oliveira, and Clark (in press) utilized both a general linear model statistic (i.e., ANOVA) as well as a random coefficient model technique to study intra- and inter-individual variability of motor movements in healthy children and children diagnosed with DCD. In summary, King and colleagues found that the random coefficient model identified intra- and inter-individual differences in task execution that the general linear model analysis did not detect. As such, the use of ANOVA to analyze intraindividual differences in automatized graphomotor performance may not have been the most appropriate or sensitive statistical model for the purposes of this research question.

**Future Research**

The results and methodology used in this study suggest multiple lines of research that should be explored in the future. First, the present study sought to understand the variability of graphomotor performance using a CV derived from the average of mean peak velocities for each word or symbol across 30 trials. Alternatively, variability could be expressed as a CV of average fluency results in both automatized and novel writing.
tasks. That is, variability of graphomotor performance in adults with ADHD could be explored using a derivative of the NJ variable rather than a derivative of mean peak velocity. In addition, and as mentioned above, future research should attempt to understand variability of graphomotor performance by utilizing an alternative calculation of the intraindividual variation of graphomotor functioning. This would involve computing the average of the CVs of mean peak velocity per trial rather than the CV of the average mean peak velocity collapsed across all trials. Finally, future research into intraindividual variability of performance could benefit from using more powerful and elegant statistical analyses. As demonstrated by King et al. (in press), the use of statistical techniques based upon a random coefficient model may have greater power to detect intraindividual differences in motor skill performance than do techniques based upon the general linear model.
APPENDICES

APPENDIX A

Interview Questionnaire

Name: ________________________________

In what month and year were you born? ______________________

How would you describe your Sex or Gender? ______________________

What hand do you primarily use to write with? Right Hand  Left Hand  Ambidextrous

What medications are you currently taking? Please include dosage information.

<table>
<thead>
<tr>
<th>Medication</th>
<th>Dosage</th>
<th>Medication</th>
<th>Dosage</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Some people use terms such as Arabic, Asian, Black, Hispanic, White, or similar terms to describe their ethnicity. What term would you use to describe your ethnicity? ______________

52
Do you currently have a diagnosis of ADHD? If yes, what is that official diagnosis?
________________________________________________________________________

Do you have a current diagnosis or diagnoses affecting the central nervous system or peripheral nervous system that would impair your ability to take part in a writing task? An example of central nervous system diagnosis affecting writing ability includes cerebral palsy affecting the arms and/or hands, and an example of peripheral nervous system diagnosis affecting writing ability includes carpal tunnel.
________________________________________________________________________

Is there any other information that you feel may affect your participation in this study that you would like me to know?
________________________________________________________________________
________________________________________________________________________
Figure B1. Scaled version of novel symbol.
REFERENCES


doi:10.1016/j.psychresns.2010.12.014


MovAlyzeR [Computer software]. Tempe, AZ: NeuroScript, LLC.


doi:10.1017/S135561779951103X

Human Movement Science, 20, 161-182. doi:10.1016/S0167-9457(01)00033-1


VITA AUCTORIS

The author, Thomas A. Duda, was born in Detroit, Michigan in January, 1977. He graduated from Algonac High School in 1995 and subsequently achieved a Bachelor of Business Administration in Accounting with Honors (Magna Cum Laude) from Baker College of Port Huron in 2000. After working in the telecommunications industry as a facility manager, systems analyst, and project manager for eight years, he returned to academic training to pursue a career in Neuropsychology and in turn obtained a Bachelor of Science in Research Psychology (Cognitive Science) with High Honors from the University of Michigan-Flint. The author was then accepted to the University of Windsor in 2010 where he later earned a Masters of Arts degree in Clinical Neuropsychology in 2012. During this time, he has conducted or contributed to research that has resulted in poster presentations at the 2012 National Academy of Neuropsychology Conference in Nashville, Tennessee; the 2010 Conference on Human Development in New York, New York; the 2010 Annual Meeting for the Child Neurology Society in Louisville, Kentucky; and the 2009 Meeting of Minds XVII Undergraduate Research Conference in Dearborn, Michigan.