

## Original Article

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# Measuring disorganized speech in schizophrenia: automated analysis explains variance in cognitive deficits beyond clinician-rated scales

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## Abstract

**Background.** Conveying information cohesively is an essential element of communication that is disrupted in schizophrenia. These disruptions are typically expressed through disorganized symptoms, which have been linked to neurocognitive, social cognitive, and metacognitive deficits. Automated analysis can objectively assess disorganization within sentences, between sentences, and across paragraphs by comparing explicit communication to a large text corpus.

**Method.** Little work in schizophrenia has tested: (1) links between disorganized symptoms measured via automated analysis and neurocognition, social cognition, or metacognition; and (2) if automated analysis explains incremental variance in cognitive processes beyond clinician-rated scales. Disorganization was measured in schizophrenia ( $n = 81$ ) with Coh-Metrix 3.0, an automated program that calculates basic and complex language indices. Trained staff also assessed neurocognition, social cognition, metacognition, and clinician-rated disorganization.

**Results.** Findings showed that all three cognitive processes were significantly associated with at least one automated index of disorganization. When automated analysis was compared with a clinician-rated scale, it accounted for significant variance in neurocognition and metacognition beyond the clinician-rated measure. When combined, these two methods explained 28–31% of the variance in neurocognition, social cognition, and metacognition.

**Conclusions.** This study illustrated how automated analysis can highlight the specific role of disorganization in neurocognition, social cognition, and metacognition. Generally, those with poor cognition also displayed more disorganization in their speech—making it difficult for listeners to process essential information needed to tie the speaker's ideas together. Our findings showcase how implementing a mixed-methods approach in schizophrenia can explain substantial variance in cognitive processes.

## Introduction

Communicating information cohesively is an essential skill when we relay instructions to others, explain the rationale behind our actions, or tell a loved one about our day. Disruptions in cohesive communication have been reported since early conceptualizations of schizophrenia – Bleuler (1911) described how many with the disorder exhibit breakdowns in connections between conscious thoughts – and are often captured by measuring disorganized symptoms. This cluster of symptoms has shown ties to deficits in neurocognition, social cognition, and metacognition (Ventura *et al.* 2010, 2013; Minor & Lysaker, 2014; Minor *et al.* 2015c). Neurocognitive deficits involve poor performance on non-social tasks in domains ranging from the attentional capacity to executive functioning (Green, 1996; Green *et al.* 2004). Social cognitive impairments reflect problems in an array of processes (e.g. theory of mind, emotion recognition) needed to draw social inferences about others (Green *et al.* 2008). Metacognitive deficits encompass difficulties integrating previous experiences to form complex ideas about oneself and others (Lysaker & Klion, 2017; Lysaker *et al.* 2018). These three cognitive areas represent distinct, but related, processes (Pinkham *et al.* 2003; Allen *et al.* 2007; Fanning *et al.* 2012; Lysaker *et al.* 2013).

In previous studies, one issue with linking disorganized symptoms to cognitive deficits has centered on the methodology implemented to measure disorganization. Typically, clinician-rated scales have been used. Although many scales are time-efficient and have been validated in schizophrenia samples (Andreasen & Grove, 1986; Bell *et al.* 1994), disadvantages of this approach include dependence on ordinal rating systems and a failure to account for how

specific facets of communication produced by examinees compare to 'typical' communication (Cohen & Elvevag, 2014; Elvevag *et al.* 2016). Research using clinician-rated scales has shown small to moderate associations between disorganized symptoms and cognitive deficits (Ventura *et al.* 2010, 2013; Hamm *et al.* 2012; Minor & Lysaker, 2014). However, since both constructs stem, in part, from breakdowns in connecting conscious thoughts, there may be additional variance that has not been identified.

One strategy to account for additional variance in cognitive deficits is to implement behaviorally-based process instruments in combination with or in place of clinician-rated scales. These instruments quantify disorganization in speech samples using either trained raters (Docherty, 2012; Docherty *et al.* 2013; Minor *et al.* 2016) or automated analysis (Elvevag *et al.* 2007; Bedi *et al.* 2015; Minor *et al.* 2015a). In both cases, specific instances of disorganization are identified and a summary score is calculated to reflect the total proportion of disorganization within each speech sample. Although the trained rater strategy benefits from a systematic approach to identify explicit behaviors, a key drawback is that the training required to produce reliable ratings and the number of hours needed to rate speech samples is time-intensive.

Automated analysis has emerged in recent years as a methodology with the potential to leverage strengths of clinician-rated and trained rater approaches (see Table 1). Using speech or writing samples, automated analysis can rapidly estimate how frequently people use certain categories of words (Buck *et al.* 2015; Minor *et al.* 2015a; Fineberg *et al.* 2016; Abplanalp *et al.* 2017; Minor *et al.* 2018) and compare explicit communication from individuals to a large corpus of text to objectively rate disorganization (Elvevag *et al.* 2007; Manschreck *et al.* 2012; Merrill *et al.* 2017; Gupta *et al.* 2018; Marggraf *et al.* 2018). In schizophrenia samples, automated analysis has been used to show how disorganization occurs across multiple levels of language, ranging from basic (e.g. words, phrases) to more complex (e.g. across sentences, paragraphs, conversations) processes (Maher *et al.* 2005; Elvevag *et al.* 2010; Moe *et al.* 2016; Willits *et al.* *in press*). Many of these studies identify disorganization at deep levels of language processing and are unlikely to be apparent using other methodologies. A recent example from Bedi and colleagues (2015) showed

how automated analysis of language structure could predict which people at clinical high risk for psychosis would convert to a psychotic disorder over a 30-month period. However, an important gap in the literature centers on comparing automated analysis with other assessments of disorganized speech to determine if automated metrics provide added value.

Implementing automated analysis alongside a clinician-rated scale of disorganization would address this gap while holding the potential to account for additional variance in cognitive processes. To date, few studies have tested relationships between disorganized symptoms measured via automated analysis and cognitive deficits, with existing studies focusing primarily on word use categories rather than disorganization (Buck *et al.* 2015; Minor *et al.* 2015a). Based on previous work, both basic and complex language processes appear to be good candidates for predicting neurocognition, social cognition, and metacognition in people with schizophrenia (Docherty *et al.* 1996; Ditman & Kuperberg, 2010; Buck & Penn, 2015; Merrill *et al.* 2017). Testing these relationships would indicate if automated analysis offers incremental variance compared with a traditional measure of disorganized symptoms. In addition, administering both a clinician-rated measure and automated analysis carries the benefit of testing how much variance in cognitive processes is explained when integrating these two methodologies.

### Objectives and hypotheses

The aims of this study were three-fold. First, we determined if disorganization measured using automated analysis was related to neurocognition, social cognition, and metacognition in schizophrenia. Our expectation was that deficits in both basic and complex language markers would be significantly linked with deficits in all three cognitive processes. Second, we tested if measuring disorganization via automated analysis added incremental variance to predictions of cognitive processes by comparing it with a clinician-rated measure. We hypothesized that automated analysis would account for significant variance when predicting neurocognition, social cognition, and metacognition and that basic and complex markers would serve as individual predictors. To compare methodologies, we also examined if the clinician-rated

**Table 1.** Comparing three methodologies of assessing speech disorganization in schizophrenia

	Clinician-rated scales	Trained raters	Automated analysis
Rater	Clinician or examiner	Trained experts	Computerized ratings
Rating of	Self-report, observations	Speech or written samples	Speech or written samples
Scale	Ordinal	Ratio	Ratio
Data collection	Examiner rates scale	Experts score samples	Automated sample scoring
Time needed	Low; rated quickly once trained on measure	High; time-intensive to train and rate	Moderate; immediate once samples transcribed
Prevalence	Widely used	Low to moderate use	Moderate but increasing
Strengths	Widely used with good psychometric data for many measures; time and cost-efficient; wide use makes it easy to compare with previous findings	Ratings based on specific behaviors; typically standardized; useful for testing contributing factors to disorganization (e.g. affect) at multiple levels of language	Fast once transcribed; produces several scales to rate behavior objectively; leverages strengths of multiple fields (e.g. psychology, computer science)
Weaknesses	Use of ordinal ratings; does not measure specific behaviors; unable to look at language components at complex levels	Time-intensive to conduct training and ratings; must check for rating drift; few longitudinal data and validation studies	Transcription takes time; number of scales can make it difficult to pinpoint measure needed or determine validation

scale accounted for variance beyond automated analysis. Finally, we determined how much variance in neurocognition, social cognition, and metacognition was explained when integrating clinician-rated and automated measures of disorganization.

## Methods

### Participants

Eighty-one outpatients from a Midwestern VA Medical Center participated in this study. All participants had: (1) a DSM-IV diagnosis of Schizophrenia ( $n = 56$ ) or Schizoaffective Disorder ( $n = 25$ ) confirmed via the Structured Clinical Interview for DSM-IV-TR Disorders-Patient Edition (SCID-I/P; First *et al.* 2002); (2) no hospitalizations or changes in medication within 30 days of testing; (3) no documented intellectual disability based on medical record review; and (4) no active substance dependence. Participants were part of a randomized controlled trial on cognitive remediation. To avoid potential confounds with a treatment designed to improve cognitive functioning, the current study focused solely on baseline testing. University and Veterans Affairs Institutional Review Boards approved all procedures and participants gave written informed consent prior to study onset.

### Measures

#### Disorganization

Automated analysis was conducted on speech generated in response to the Indiana Psychiatric Illness Interview (IPII; Lysaker *et al.* 2002), a semi-structured interview that assesses perceptions of one's life and illness. The open-ended nature of the IPII was a key reason for its selection; its format differs from many structural interviews and speech tasks in that subjects control how long they speak with little input or affective prompting from examiners. IPII interviews were typically 30–60 min in length, allowing subjects to generate substantial samples for analysis (Total words: Mean = 2786, *s.d.* = 2117). Each interview was recorded and transcribed with examiner speech removed.

Automated analysis of IPII interviews was conducted using Coh-Metrix 3.0 (McNamara *et al.* 2014). Coh-Metrix is a discourse processing software that analyzes language by measuring coherence and cohesion, which are closely tied to disorganization (Andreasen, 1979; Elvevag *et al.* 2007). It has been used in several studies (Crossley *et al.* 2009; McCarthy & Jarvis, 2010; see

McNamara *et al.* 2014) and has shown the ability to differentiate high and low levels of organization in healthy adults (McNamara *et al.* 2006) and people with schizophrenia (Willits *et al.* in press). Coh-Metrix contains 108 indices across 11 categories and provides a comprehensive view of language characteristics ranging from basic to complex processes. Basic indices typically examine speech at word or phrase levels, whereas complex indices measure organization within sentences, between sentences, and across paragraphs.

For this study, we selected two basic (syllables per word, type-token ratio) and five complex indices (narrativity, syntactic simplicity, word concreteness, referential cohesion, deep cohesion; see Table 2 for descriptions of all indices) a priori. Basic indices were chosen based on their expected overlap with cognitive processes. Specifically, syllables per word corresponds with verbal intelligence which is inversely linked to disorganization (O'Leary *et al.* 2000; Ventura *et al.* 2010); type-token ratio has been used as a disorganization marker (Manschreck *et al.* 1981, 1991; Allen, 1983) and is a key component of systems measuring language (e.g. CAST; Maher *et al.* 2005). Complex indices were selected due to their ability to capture discourse characteristics across full speech samples. The specific complex indices chosen here were narrowed down based, in part, on a principal component analysis study conducted by co-creators of Coh-Metrix 3.0 that showed these five indices accounted for the majority of variance in 54 separate categories (Graesser *et al.* 2011). To date, Coh-Metrix has only been used to analyze speech in one schizophrenia study (Willits *et al.* in press) and written text in one study of people at clinical high risk for psychosis (Gupta *et al.* 2018).

The disorganized symptoms subscale of the Positive and Negative Syndrome Scale (PANSS; Kay *et al.* 1987; see Bell *et al.* 1994) was used as the clinician-rated measure in this study. PANSS items range from 1 (absent) to 7 (extreme) and have been used extensively to rate symptoms in people with schizophrenia (Kay *et al.* 1987; Bell *et al.* 1992; Minor *et al.* 2015c). Seven items comprise the disorganized subscale: difficulty in abstract thinking, stereotyped thinking, conceptual disorganization, lack of insight, poor attention, tension, and mannerisms/posturing. Our sample demonstrated moderate disorganization with considerable range between participants ( $M = 16.63$ , *s.d.* = 3.75, range: 11–27). The PANSS has shown strong psychometric properties (Kay *et al.* 1987; Bell *et al.* 1992; Lysaker *et al.* 2013); it was rated by clinicians who had undergone several weeks of PANSS training. Training consisted of completing sample ratings on videotapes of clinical interviews. Before administering the

**Table 2.** Means, standard deviations, and descriptions of automated analysis indices ( $n = 81$ )

Basic indices	What index measures	<i>M</i>	<i>s.d.</i>	Range
Syllables per word	Complex word knowledge by analyzing mean syllables per word spoken	1.27	0.04	1.19 to 1.37
Type-token ratio	Total number of different words (i.e. type) divided by total words (i.e. token)	0.23	0.08	0.09 to 0.58
Complex indices				
Narrativity	If speech involves previously introduced characters, places, or topics to tell a story	2.44	0.41	0.90 to 3.19
Syntactic simplicity	How easy sentences are to process and if they have simple syntactic structures	−0.12	0.70	−2.76 to 1.50
Word concreteness	If content words are easy to process and represent visually rather than abstract	−1.16	0.66	−3.50 to 0.04
Referential cohesion	How words and ideas connect across sentences and throughout conversations	1.34	0.64	0.16 to 3.28
Deep cohesion	If causal and logical links are present to help others form deeper understanding	0.46	0.62	−0.81 to 2.00

Notes. Multidimensional indices reflect *z*-scores from Coh-Metrix 3.0. Additional information on all indices available in McNamara *et al.* (2014).

PANSS, clinician's ratings had to demonstrate good inter-rater reliability ( $\alpha = 0.80$ ) with a consensus rating group with several years of experience using the PANSS.

### Neurocognition

The Measurement and Treatment Research to Improve Cognition in Schizophrenia (MATRICS; Nuechterlein *et al.* 2008), which was designed specifically to assess neurocognitive functioning in schizophrenia (De Herdt *et al.* 2013; McCleery *et al.* 2015; Minor *et al.* 2015b), was used as the neurocognition measure. In this study, a composite neurocognition score was created for each participant using six of the seven MATRICS domains (processing speed, attention, working memory, verbal learning, visual learning, reasoning, and problem-solving). The MATRICS social cognition measure, the Mayer-Salovey-Caruso Emotional Intelligence Test (MSCEIT; Mayer *et al.* 2002), was excluded from this score and included with social cognition. Equal weight was given to the remaining six domains.

### Social cognition

A composite score was created for social cognition that gave equal weight to three separate domains: (1) emotional processing (measured using the MSCEIT); (2) emotion recognition [Bell-Lysaker Emotion Recognition Task (BLERT); Bell *et al.* 1997]; and (3) theory of mind [Hinting task (Corcoran *et al.*, 1995; Greig *et al.* 2004) and Social Attributions Test-Multiple Choice (SAT-MC; Bell *et al.* 2010)]. The MSCEIT (Nuechterlein *et al.*, 2008; Eack *et al.*, 2010), BLERT (Bell *et al.* 1997), Hinting task (Bell *et al.* 2010), and SAT-MC (Pinkham *et al.* 2014) have all demonstrated strong psychometric properties in schizophrenia; the BLERT and Hinting task were selected by the Social Cognition Psychometric Evaluation (SCOPE) study as being among the best social cognition measures in schizophrenia (Pinkham *et al.* 2014).

### Metacognition

The Metacognition Assessment Scale-Adapted (MAS-A; Lysaker *et al.* 2005, adapted from Semerari *et al.* 2003) was used to assess metacognition. The MAS-A assesses a person's ability to integrate implicit and explicit information about themselves and others to form an integrated sense of how to respond to psychosocial challenges. The MAS-A is scored from 0 to 28, with higher scores signifying greater metacognitive capacity (Lysaker *et al.* 2013). MAS-A scores were calculated using raters who had exhibited good inter-rater reliability ( $\alpha \geq 0.80$ ). In schizophrenia cohorts, the MAS-A has demonstrated good reliability and validity in several studies (see Lysaker & Dimaggio, 2014).

### Analysis

Data analysis was conducted in three parts. First, composite scores were calculated for neurocognition and social cognition. For neurocognition, standardized *t*-scores were converted to *z*-scores and summed using a weighted average of 1.0 for all MATRICS domains except social cognition. For social cognition, measures were converted to *z*-scores and summed, with the MSCEIT (emotion processing) and BLERT (emotion recognition) given weighted averages of 1.0 and the Hinting task and SAT-MC (theory of mind) given weighted averages of 0.5. This strategy gave equal weight to each social cognition domain. Second, correlations between cognitive variables (neurocognition, social cognition, metacognition) and Coh-Metrix indices were conducted to test relationships between cognitive processes and disorganization.

Correlations between disorganization measures were also analyzed. Finally, two sets of stepwise regressions were run with neurocognition (regression one), social cognition (two), and metacognition (three) as outcome variables. In the first set, PANSS disorganized symptoms were entered as the step one predictor, and Coh-Metrix indices as step two predictors. The order was reversed in the second set. For all regressions, only those Coh-Metrix indices that exhibited a significant relationship with cognitive variables in correlational analyses were entered.

### Results

Participants were primarily middle aged ( $M = 49.74$ ,  $S.D. = 10.71$ ), male ( $n = 77$ , 95%), African-American ( $n = 46$ , 57%), currently unmarried ( $n = 47$ , 58%), unemployed ( $n = 68$ , 84%), reported only immediate family or providers in their social network ( $n = 53$ , 65%), completed high school/GED ( $n = 74$ , 91%), and were earning below \$20 000 annually ( $n = 69$ , 85%). On average, they were first hospitalized in their late twenties ( $M = 29.43$ ,  $S.D. = 12.35$ ) and were currently prescribed antipsychotic medications (Chlorpromazine equivalent  $M = 363.64$ ,  $S.D. = 416.10$ ). Post-hoc analyses of associations between disorganization methods showed small convergence between the clinician-rated scale with referential cohesion,  $r(79) = -0.24$ ,  $p = 0.029$ , syllables per word,  $r(79) = -0.12$ ,  $p = 0.298$ , type-token ratio,  $r(79) = -0.21$ ,  $p = 0.067$ , syntactic simplicity,  $r(79) = 0.20$ ,  $p = 0.079$ , word concreteness,  $r(79) = -0.11$ ,  $p = 0.312$ , and deep cohesion,  $r(79) = -0.11$ ,  $p = 0.324$ . Medium convergence was observed with narrativity,  $r(79) = -0.31$ ,  $p = 0.005$ .

### Links between cognitive processes and disorganization using automated analysis

As shown in Table 3, several small to medium significant relationships were observed between cognitive processes and disorganization measured via automated analysis. Neurocognition was associated with one basic index (syllables per word) and inversely associated with one complex index (syntactic simplicity). Social cognition was inversely associated with one complex index (syntactic simplicity) but was not linked with basic indices. Metacognition was inversely associated with one basic index (type-token ratio) and positively related to one complex index (referential cohesion). Three complex indices (narrativity, word concreteness, and deep cohesion) were not significantly associated with any cognitive process. These findings generally supported hypotheses that basic and complex indices of disorganization would be related to neurocognition, social cognition, and metacognition.

### Variance in cognitive processes explained by automated analysis when accounting for a clinician-rated measure of disorganized symptoms

To determine if automated analysis explained added variance in cognitive processes, stepwise regression models were conducted with a clinician-rated scale of disorganized symptoms entered as the step one predictor. In the first model, one basic (syllables per word) and one complex (syntactic simplicity) automated index were entered simultaneously as step two predictors of neurocognition. The overall model was significant,  $F_{(3,77)} = 9.73$ ,  $p < 0.001$ , with clinician-rated and automated measures of disorganization accounting for 28% of the variance in neurocognition (adjusted  $R^2 = 0.25$ ). Automated indices explained 10% of the variance beyond the clinician-rated scale. All three variables

**Table 3.** Relationships between disorganization using automated analysis with neurocognition, social cognition, and metacognition ( $n = 81$ )

	Neurocognition	Social Cognition	Metacognition
Descriptive indices			
Syllables per word	0.25*	0.13	-0.15
Type-token ratio	-0.16	-0.09	-0.46***
Multidimensional indices			
Narrativity	0.01	0.10	0.18
Syntactic simplicity	-0.30**	-0.28*	-0.14
Word concreteness	0.08	0.01	0.15
Referential cohesion	0.13	0.13	0.26*
Deep cohesion	0.12	0.20 <sup>+</sup>	0.19 <sup>+</sup>

\*\*\* $p < 0.001$ ; \*\* $p < 0.01$ ; \* $p < 0.05$ ; <sup>+</sup> $p < 0.10$ .

entered (clinician-rated disorganization, syllables per word, syntactic simplicity) were significant predictors of neurocognition. For social cognition, one complex index (syntactic simplicity) was entered into the second step of the regression. The overall model was significant,  $F_{(2,78)} = 15.05$ ,  $p < 0.001$ , accounting for 28% of the variance in social cognition (adjusted  $R^2 = 0.26$ ). Clinician-rated disorganization was the lone significant predictor. The automated index (syntactic simplicity) was on the cusp of significance and contributed 3% of variance beyond the clinician-rated scale. In the third model, metacognition was the outcome

variable; one basic (type-token ratio) and one complex (referential cohesion) index were entered as predictors. The overall model was significant,  $F_{(3,77)} = 11.40$ ,  $p < 0.001$ , accounting for 31% of the variance in metacognition (adjusted  $R^2 = 0.28$ ). Automated indices explained 26% of the variance after accounting for the clinician-rated scale. Both clinician-rated disorganization and the basic automated index (type-token ratio) were significant individual predictors. These findings generally supported our hypothesis that disorganization measured via automated analysis could explain added variance in cognitive processes (Table 4).

**Table 4.** Stepwise regressions using clinician-rated (step one) and automated measures (step two) of disorganized symptoms as predictors of cognitive variables in schizophrenia ( $n = 81$ )

	$R^2$	$B$	s.e. $B$	$\beta$
<i>Model one (Neurocognition)</i>				
Step one	0.18***			
Disorganized symptoms		-0.47	0.11	-0.42***
Step two	0.28***			
Disorganized symptoms		-0.38	0.11	-0.35**
Syllables per word		25.54	11.39	0.22*
Syntactic simplicity		-1.46	0.59	-0.24*
<i>Model two (social cognition)</i>				
Step one	0.25***			
Disorganized symptoms		-0.30	0.06	-0.50***
Step two	0.28***			
Disorganized symptoms		-0.27	0.06	-0.46***
Syntactic simplicity		-0.59	0.31	-0.19 <sup>+</sup>
<i>Model three (metacognition)</i>				
Step one	0.05 <sup>+</sup>			
Disorganized symptoms		-0.06	0.03	-0.21 <sup>+</sup>
Step two	0.31***			
Disorganized symptoms		-0.09	0.03	-0.32**
Type-token ratio		-6.69	1.37	-0.53***
Referential cohesion		-0.02	0.17	-0.01

\*\*\* $p < 0.001$ ; \*\* $p < 0.01$ ; \* $p < 0.05$ ; <sup>+</sup> $p < 0.10$ .

### Variance in cognitive processes explained when reversing stepwise regressions

Stepwise regression models were also reversed with automated analysis indices entered in step one and the clinician-rated scale entered in step two (Table 5). For neurocognition, the clinician-rated scale accounted for 12% of the variance after controlling for automated indices. The clinician-rated scale accounted for the most variance, 20%, in social cognition and the least variance, 9%, in metacognition whenever automated indices were controlled for. These findings illustrate the unique variance in cognitive processes accounted for by a commonly-used clinician-rated scale when controlling for automated indices.

### Discussion

Our primary goals in this study were to use automated analysis to delineate links between disorganization and cognitive processes and determine if automated analysis added incremental variance in predicting cognitive processes compared with a commonly used clinician-rated measure. Three key findings emerged. First, complex automated indices were significantly associated with all three cognitive processes and basic indices were related to neurocognition and metacognition, but not social cognition. Second, automated analysis explained significant variance in neurocognition and

metacognition even when accounting for the clinician-rated scale. Third, automated analysis and the clinician-rated instrument combined to account for between a quarter to a third of the total variance in all three cognitive processes.

When assessing disorganized symptoms broadly, previous research has shown these symptoms are related to cognitive processes (Ventura *et al.* 2010, 2013; Hamm *et al.* 2012; Minor & Lysaker, 2014); the current study is novel in its implementation of automated analysis to illustrate which specific facets are linked with neurocognition, social cognition, and metacognition. Regarding basic language indices, findings indicated that automated analysis reveals relationships between neurocognition and metacognition with disorganization at fundamental levels of discourse. The most intriguing finding was that people with poor metacognitive capacity produced a greater frequency of different words (i.e. higher type-token ratio). Although counterintuitive at first glance, this may reflect the inverse relationship between type-token ratio and a number of words spoken (McCarthy & Jarvis, 2010), as people with lower metacognitive capacity tended to generate less speech in clinical interviews (Buck *et al.* 2015). A post-hoc analysis testing this explanation showed that number of words spoken was significantly associated with metacognition—accounting for 14% of the variance—and inversely related to type-token ratio—accounting for 55% of the variance. A second possibility is that those with poor metacognitive capacity

**Table 5.** Reverse stepwise regressions entering automated (step one) before clinician-rated measures (step two) to predict cognitive variables in schizophrenia ( $n = 81$ )

	$R^2$	$B$	s.e. $B$	$\beta$
<i>Model one (neurocognition)</i>				
Step one	0.16**			
Syllables per word		30.57	12.07	0.26*
Syntactic simplicity		-1.87	0.62	-0.31**
Step two	0.28***			
Syllables per word		25.54	11.39	-0.22*
Syntactic simplicity		-1.46	0.59	-0.24*
Disorganized symptoms		-0.38	0.11	-0.35**
<i>Model two (social cognition)</i>				
Step one	0.08***			
Syntactic simplicity		-0.88	0.35	-0.28*
Step two	0.28***			
Syntactic simplicity		-0.59	0.31	-0.19†
Disorganized symptoms		-0.27	0.06	-0.46***
<i>Model three (metacognition)</i>				
Step one	0.22***			
Type-token ratio		-5.34	1.38	-0.42***
Referential cohesion		0.15	0.17	0.09
Step two	0.31***			
Type-token ratio		-6.69	1.37	-0.53***
Referential cohesion		-0.02	0.17	-0.01
Disorganized symptoms		-0.09	0.03	-0.32**

\*\*\* $p < 0.001$ ; \*\* $p < 0.01$ ; \* $p < 0.05$ ; † $p < 0.10$ .

frequently shifted between topics—leading to more unique words—without providing the connective threads necessary to frame the context for their audience. This is in line with the observed association between metacognition and referential cohesion found here and supports findings of greater disorganization in those with schizophrenia who display poor metacognition (Hamm *et al.* 2012; Minor & Lysaker, 2014).

Neurocognition, social cognition, and metacognition were each associated with at least one complex automated index, suggesting that these different cognitive abilities are connected to disorganization at sophisticated levels – within sentences, between sentences, and across paragraphs – of discourse. Associations were in the small to medium effect size range. In all instances, those who performed poorly in a cognitive area also produced speech that lacked clarity, making it challenging for listeners to devise meaning. Specifically, automated analysis captured how those with poor neurocognition and social cognition used opaque, unfamiliar sentence structures. This style forces listeners to work to understand the speaker's intentions, creating obstacles when processing information. The automated analysis also showed how people with poor metacognitive capacity produced speech containing few overlapping words and ideas; thus, leaving out the important context for listeners by omitting explicit threads to connect concepts. Without these connective threads, the audience is typically unable to integrate essential information to tie the speaker's ideas together. Although it was surprising some complex indices (narrativity, word concreteness, deep cohesion) were not related to cognitive processes, our general findings supported previous observations of links between speech disorganization and cognitive impairment (Ventura *et al.* 2010, 2013; Minor & Lysaker, 2014).

The size of observed associations using automated analysis was generally in line with previous studies implementing clinician-rated or trained rater approaches. Clinician-rated scales of disorganization have demonstrated small (neurocognition; Ventura *et al.* 2013) or small to medium effect size associations (social cognition; Ventura *et al.* 2013; metacognition; Hamm *et al.* 2012; Minor & Lysaker, 2014) with cognitive processes. The trained rater approach has yielded small to medium associations with neurocognitive and social cognitive variables (Docherty, 2012; Docherty *et al.* 2013), typically outperforming clinician-rated scales as well as automated analysis based on the links observed in this study. In line with the relatively small associations observed here, the trained rater approach has also demonstrated only minor or modest convergence with clinician-rated scales (Docherty *et al.* 1997; Docherty, 2005, 2012). Comparing automated and trained rater methods is an important future avenue of research.

Going beyond associations, one of the most compelling aspects of this study is that it determined how much unique variance automated analysis explained in cognitive processes beyond a traditional measure of disorganized symptoms. Estimates of neurocognition, social cognition, and metacognition varied widely. On the low end, automated analysis minimally explained social cognition (3%) and no automated index was a significant predictor. In contrast, automated analysis had a moderate role in neurocognition, explaining 10% of the variance beyond the clinician-rated measure. Both syllables per word (basic index) and syntactic simplicity (complex index) individually predicted neurocognition. This highlights how people with schizophrenia displaying poor neurocognition produce simple words that should be easy for others to comprehend, yet their sentences contain

unfamiliar structures that are difficult for their audience to follow. Of the cognitive processes tested here, automated analysis explained the most unique variance in metacognition (26%). Type-token ratio (basic index) was a significant individual predictor, indicating that those with schizophrenia who used a greater frequency of different words tended to display poor metacognitive capacity even after accounting for a clinician-rated scale.

When controlling for automated analysis, the clinician-rated scale accounted for significant variance in neurocognition (12%), social cognition (20%), and metacognition (9%). Compared with automated analysis, the clinician-rated scale accounted for more unique variance in social cognition (20% *v.* 3%), marginally more variance in neurocognition (12% *v.* 10%), and considerably less variance in metacognition (9% *v.* 26%). One important difference between these two methods is that a composite score of seven clinician-rated disorganized symptoms was tested compared with only one or two automated indices (which were selected based on associations with the cognitive process). Taken together, our findings showcase the unique variance accounted for by two different methodologies and illustrate the value of implementing each method for explaining variance in different cognitive processes.

Whereas automated analysis and the clinician-rated scale accounted for a wide range of variance in cognitive processes alone, the amount explained whenever they were combined was more consistent – between 28 and 31%. Both automated analysis and clinician-rated scales assess disruptions in conscious thoughts, with the former using computational software to characterize disorganized symptoms at basic and sophisticated levels of discourse (McNamara *et al.* 2014) and the latter relying on clinician's impressions of a person's level of disorganized speech and behavior (Kay *et al.* 1987). A key contribution of this study is the observation that integrating these approaches is useful for illustrating the role of disorganization in neurocognition, social cognition, and metacognition. Given that the study sample exhibited mild to moderate disorganization, a novel future direction would be to select subjects based on the presence of formal thought disorder and determine if a mixed-methods approach accounts for additional variance in cognitive processes. This strategy would be compatible with calls from funding agencies (Insel, 2014) and researchers in the field (see Cohen *et al.* 2017).

Study strengths include using novel automated indices to calculate objective assessments of disorganization and comparing automated analysis to traditional measurements. One limitation is that data were only collected at one time point. This is particularly relevant for speech samples, given that Coh-Metrix has only been used in one previous schizophrenia study (Willits *et al.* *in press*) and reproducibility is a critical issue when testing new methodologies. Our group is currently collecting longitudinal data to measure the test-retest reliability of Coh-Metrix indices on several types of speech data, including IPII interviews. A second limitation is that the clinician-rated scale contained items that may not be associated with language dysfunction (e.g. lack of insight). Although previous studies have shown that the seven items used here cluster together (Bell *et al.* 1994), future studies may benefit from implementing scales where all items are designed to assess behavioral disorganization. A third limitation is that no control group was recruited to compare how automated analysis predicted cognitive functioning in healthy adults. In a previous study, Coh-Metrix indices did differentiate schizophrenia and control groups (Willits *et al.* *in press*); however, this study did not test relationships between disorganization and cognitive processes. A fourth limitation concerns generalizability; our

sample tended to be better educated, had fewer females, and were hospitalized later in life than many schizophrenia cohorts. Although these characteristics are common in VA samples (Harvey *et al.* 2000; Thorp *et al.* 2012; Firmin *et al.* 2016), future work should replicate our findings using community subjects.

In sum, this is among the first studies showing how automated analysis can be implemented to predict cognitive processes in schizophrenia. Significant associations were observed between automated indices and neurocognition, social cognition, and metacognition. When accounting for a traditional clinician-rated measure, automated analysis varied in its ability to predict cognitive processes: whereas little variance was accounted for in social cognition, substantial variance was explained in metacognition and multiple indices were significant predictors of neurocognition. Finally, integrating automated analysis with a clinician-rated measure explained between a quarter to a third of the variance in all three cognitive processes. This finding highlights the benefits of implementing a mixed-methods approach. Future studies should replicate these results using longitudinal data and determine how disorganization predicts cognitive processes in community and formal thought disorder samples.

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**Ethical standards.** The authors assert that all procedures contributing to this work comply with the ethical standards of the relevant national and institutional committees on human experimentation and with the Helsinki Declaration of 1975, as revised in 2008.

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