Objective and Rationale

Treatment evaluation and research requires discourse measures that are: VALID, RELIABLE, and RELEVANT but also PRACTICAL in terms of labor and training.

Primary limiting factors for linguistic discourse analysis are: TIME and CLINICAL KNOWLEDGE¹.

This project describes an automatic system for measuring grammatical complexity in discourse using CLAN².

The system uses a grammatical relations (GR) parser¹ that has recently been trained on adult language samples and tested for computation of a syntactic complexity index.

By identifying the GRs that mark embedding, a Grammatical Relations-Complex (GR-C) measure can be calculated.

This GR-C measure adds important and relevant linguistic data that can be computed on multiple samples with accuracy, replicability, speed, and flexibility.

Background

Linguistic analyses in aphasia often focus on grammatical aspects^{3,4} because syntax is a key component of aphasia diagnosis and treatment.

Many measures can be computed automatically – e.g., total utterances, total words, total unique words, TTR, MLU, words per minute, frequencies of parts-of-speech, morphological affixes, proposition density, repetitions, revisions. However, grammatical complexity has been less amenable to automatic computation.

Embedding, or recursion, is a structural indicator considered to be a good indicator of syntactic complexity. Systems used more commonly for child language, such as LARSP⁵, use embeddings to compute grammatical complexity.

GRASP (Grammatical Relations Analyzer for Spontaneous Protocols) is a parser developed to accurately and automatically measure syntactic complexity by producing a tier for grammatical relations (GRs) in CHAT files⁶.

Recent training of the GRASP parser on adult language samples allows us to compute grammatical complexity from the GR codes.

All of the GRs (n=45) are explained in detail in the CLAN Manual http://talkbank.org/manuals/MOR.pdf)

Running the MOR command in CLAN automatically produces a %mor tier with morphosyntactic analysis and a %gra tier with pairwise GRs words in a matter of seconds. For example:

*PAR: the big dog chased five cats.

%mor: det the adj big n dog v chase-PAST quant five n cat-PL %gra: 1|3|DET 2|3|MOD 3|4|SUBJ 4|0|ROOT 5|6|QUANT 6|4|OBJ

7|4|PUNCT

In this sentence:

- the DETerminer (the) and MODifier (big) attach to the SUBJect (dog) which attaches to the verb (chased), the ROOT of the clause
- the QUANTifier (five) attaches to OBJect (cats) which attaches to the verb.

The numbers indicate the pairwise grammatical relations, where the first number is the word's order in the sentence and the second number indicates its attachment.

The %gra tier can also be visually represented:

SUBJ (chase) ncat-PL MOD QUANT DET PUNCT the big dog chased five cats .

Testing this automated GRASP system with human coding of embeddings in adult discourse samples yielded an overall accuracy of 95% (of the 74 embeddings spotted by the automated system, two were false alarms and one was missed).

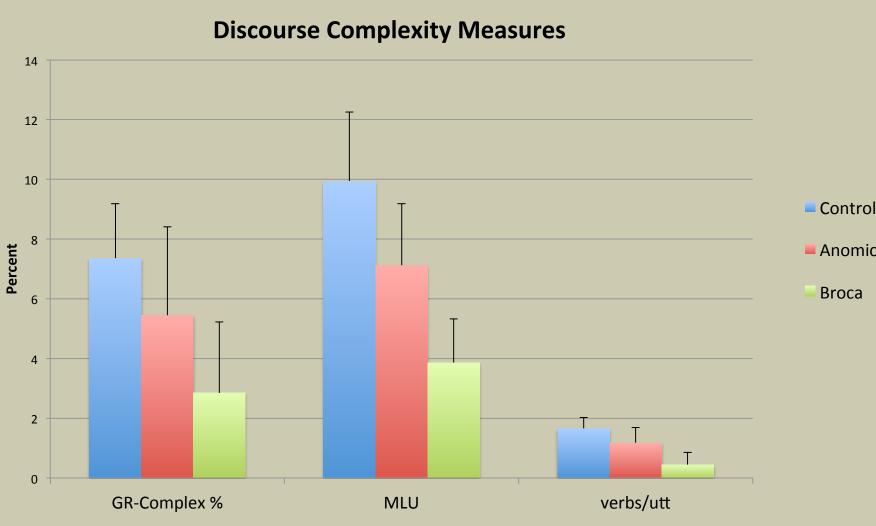
Automatic Grammatical Complexity Analysis in Aphasia

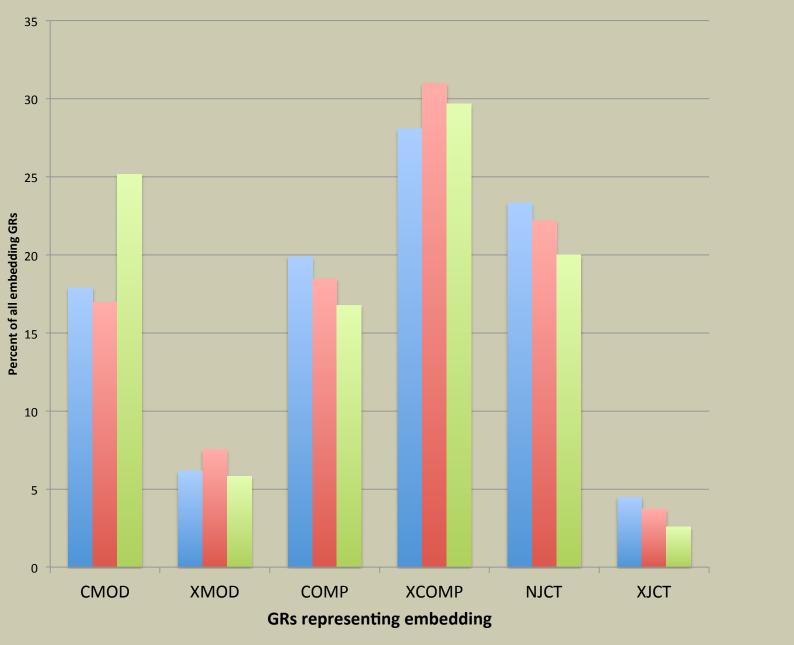
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Resea	rch Questic	ns				Results			
Will the Grammatical Relations-Complex (GR-C) measure reveal differences between Cinderella narratives produces by people with nonfluent aphasia (Broca's), fluent aphasia (Anomic), and people without aphasia?						 All three groups were significantly different on GR-Complex [F(2, 325) = 85.1, p = .000)]. 			
Does the GR-C index correlate with other measures used as indicators of grammatical complexity – MLU and # of verbs/utterance?						 Post-hoc testing (Bonferroni multiple comparison test) revealed significant GR-Complex differences between all groups with the Broca's group having the lowest GR-Complex score (2.85%) and the non-aphasic group having the highest (7.36%). 			
Meth	ods								
Particip						 Similar results were 173.5, p = .000)] and 		•	
•		databa	nse ⁷ , inclusion cr	iteria were:					
• apha	e English speak sia as a result o	f stroke				 Correlations between GR-Complex and # of verbs per utterance and MLU were strong, positive, and significant (p < .01). Discourse data – means and standard deviations 			
	words on the C session only, if _l		la task ant was seen m	ultiple times					
Demog	raphic characte	ristics -	- means and sta	ndard deviations			Anomic (n=87)	Broca (n=50)	Controls (n=160)
		iomic =87)	Broca (n=50)	Controls (n=191)		GR-Comple x**	5.45 % (2.96)	2.85 % (2.38)	7.36 % (1.82)
Age	(yrs.) 62.3	(12.2)	56.8 (11.3)	62.0 (19.2)					
Educati	on (yrs.) 15.8	(2.8)	14.5 (2.4)	15.4 (2.4)		MLU**	7.1 (2.1)	3.9 (1.5)	9.9 (2.3)
	,	8%	54%	45%		# verbs/utterance**	1.2	0.5	1.7
WAB	-R AQ 85.0	(6.5)	57.0 (9.7)	NA			(0.5)	(0.4)	(0.4)
No sign	ificant group di	fference	es on age, sex, e	education.		total words**	226.3 (174.6)	97.3 (69.6)	473.3 (282.3)
Utteran	iptions were co ces were segme	ented u	sing the QPA hi	d experienced transcri erarchy ⁸ : syntax, inton anscription and reache	ation, pause,	total utterances*	31.1 (21.6)	25.54 (17.2)	48.7 (34.1)
	agreement on a e at <u>http://apha</u>	•	•	plete transcripts and v	videos are	** significant difference Repformani tost for		• • • •	according to
1. Run	MOR on all tra	nscripts	s → %	6 mor and %gra tiers		Bonferroni test forsignificant difference			ips and control g
				spread al number of GRs)		(p<.01) but not bety Bonferroni test for i		—	ps, according to
Of the 41 possible grammatical relations (excluding 4 cosmetic punctuation markers), the 10 that mark syntactic embeddings are:					ctuation	Discourse Complexity Measures			
COMP	the finite clau <i>I think that we</i>		nplement of a ve er.	erb		12			
XCOMP			l complement o <i>a can of tuna.</i>	f a verb		Bercent 8			 ,
CPRED	a full clause t My goal is to		•	cate nominal of verbs				Ţ	
CPOBJ	a full clause t I'm not clear d		-	of a preposition		2			T
COBJ	a full clause tl I remember w		ves as the direct a said .	object		GR-Complex %	MLU		erbs/utt
CSUBJ	the finite clau <i>That Eric cried</i>	•	ject of another o <i>I Bush.</i>	clause		% Usage	of Individual Emb	edding GRs	
XJCT			nat attaches to a <i>iting museums d</i>	verb, adjective, or ad and galleries.	verb	30			
NJCT	attached as ar	adjund	•	vith a prepositional ph volver.	rase	25 89 10 20			
CMOD	a finite clause He was happy			ifier or complement		be all emperations of all empera			
XMOD	a non-finite c <i>It's time to ta</i>			modifier or compleme	ent	10			
-			•	UBJ (subject), OBJ (obj	· · ·	5			
object),	JCT (adjunct),	MOD (n	nodifier), DET (d	ve), POBJ (preposition eterminer), QUANT), NEG (negation), INF		0 CMOD XMOI	D COMP XC	OMP NJCT	XJCT

(quantifier), POSS (possessive), AUX (auxiliary), NEG (negation), INF (infinitive), COM (communicator).

group





Results, cont.

Pear

Conclusions and Future Directions

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Control

Anomic

Broca

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on pro	product moment correlations						
		GR-Complex	MLU				
	MLU	.65*					
	# verbs/utterance	.67*	.93*				

* significant correlation coefficients (p <. 01)

• The GR-complex measure is a promising and practical tool for automatic analysis of syntax in clinical and research discourse analyses.

• Once a discourse sample is transcribed in CHAT format, this complexity index can be automatically calculated with no further coding or annotation.

• These results are consistent with the literature, showing more embeddings in language samples of non-aphasic participants than PWA⁹ and more general findings of reduced syntactic complexity in Broca's aphasia¹⁰.

• Use the GR-complex measure:

1. to evaluate the effect of syntax treatment programs on discourse;

2. with other populations such as Primary Progressive Aphasia, where recent neuroimaging results showed that reduced frequency of embeddings was associated with atrophy in the left frontal lobe (posterior inferior frontal gyrus, superior frontal sulcus and adjacent prefrontal regions and the supplementary motor area)¹¹; and

3. to learn more about the types of embeddings used in different discourse genres and in participants with different types and severities of aphasia.

Ideas for improving the GR measure include the following:

• Address the fact that CSUBJ and COBJ were not coded properly and CPRED and COBJ were coded too infrequently on the %gra tier;

• Consider the new Google Universal Dependency Relations initiative and consider using the relativizer as the head of the subordinate clause and simplifying the %gra tier coding (use SUBJ and OBJ instead of CSUBJ and COBJ) and do cross-tier searches (for example, for relativizers on the %mor tier linked to the verb as SUBJ on the %gra tier);

• Do additional training of the %gra tier with more discourse genres and more samples.

³ Berndt, R.S., Wayland, S., Rochon, E., Saffran, E., & Schwartz, M. (2000). Quantitative production analysis (QPA). Philadelphia: Psychology Press.

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