

Overview

We present a computer program, CPIDR, for measuring propositional idea density in English text

Propositional idea density is the amount of information (as a proposition count) divided by the number of words. It is a fundamental measurement in the study of discourse comprehension and is also a clinically useful diagnostic indicator.

Snowdon et al. (1996) found reduced idea density in the writing of Alzheimer's Disease victims 50 years before the onset of symptoms.

CPIDR (Computerized Propositional Idea Density Rater, pronounced "spider") counts propositions by counting verbs, adjectives, adverbs, prepositions, and conjunctions (as suggested by Snowdon et al. 1996). It then applies adjustment rules to make the count more accurate.

Propositions

A proposition is a piece of information, an idea or belief that can be true or false.

Proposition theory was introduced by Kintsch (1974) and has become a standard part of the methodology of applied psycholinguistics.

Unlike later formal semantics, Kintsch's theory does not count common nouns as propositions (predicates), nor does it count verb tense or modality separately from the verb itself.

Propositions in a text are normally identified by trained human raters following the handbook of Turner and Greene (1977) and measurement is subject to local and personal variation. We present a technique for counting propositions objectively through part-of-speech tagging.

CPIDR uses the Java edition of MontyTagger (Liu, 2004) to tokenize the input text and tag the parts of speech. MontyTagger is based on the earlier Brill tagger and the Penn Treebank.

Then CPIDR adjusts its proposition count by applying a set of adjustment rules. For example, a linking verb is not counted separately from its adjective; *seems old* is counted as one proposition, not two.

CPIDR's adjustment rules are implemented as a Java program that scans the tagger's output using a 4-item moving window, deciding whether to count the last item of the four. If it is discovered that an earlier item should not have been counted, its count is decremented.

The final output consists of the proposition count, the word count, and the quotient of the two (idea density).

The adjustment rules in CPIDR were constructed to handle all the example sentences in Turner and Greene (1977) and obvious generalizations of them.

Count a verb in the set *is, seems, looks, smells, becomes...* followed by an adjective or adverb as a linking verb. (Actual set is larger.)

(Etc., for a total of about 20 rules)

Measuring Propositional Idea Density through Part-of-Speech Tagging

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How CPIDR Works

Basic Algorithm

Proposition Count =

+ Adjectives

Verbs

- + Adverbs
- + Prepositions
- + Conjunctions
- + Determiners (except *a*, *an*, *the*)
- Modals (only if negative)
- Auxiliary verbs
- Linking verbs

Adjustment Rules

(examples)

Do not count *the, a,* or *an* as a determiner.

Count a verb immediately followed by *not* as an auxiliary verb.

Count either...or as one conjunction, not two.

Do not count modals unless they end in *n't*.

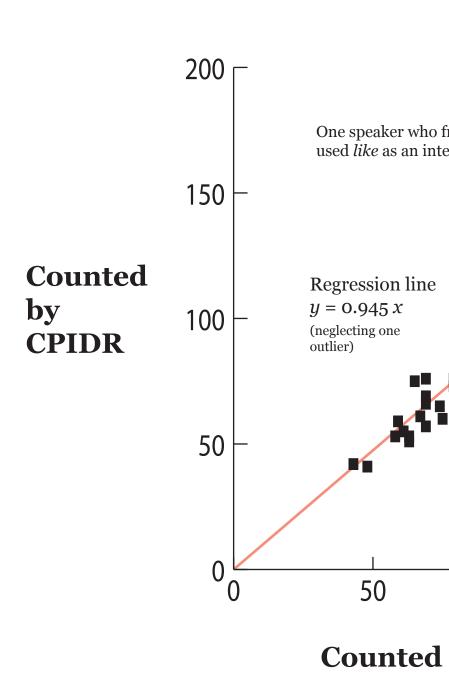
Example

Sentence:

Testing CPIDR CPIDR was tested on 40 samples of spontaneous speech previously collected and analyzed into I called my dad because I was scared. propositions by co-authors Kemper and Herman. Language samples were elicited from 40 (1) called(I,dad) volunteers in response to the question, "What do (2) my(dad) you remember about the morning of 9/11?" (3) because(1,4)(4) scared(I) The final 10 sentences of each sample were analyzed for grammatical complexity as detailed eliminary Adjusted by Kemper et al. (1989), then analyzed into proposition oposition propositions following Turner and Greene unt = 5count = 4 (1977). Five different human raters analyzed propositions; their agreement exceeded r = 0.82. called led my These transcripts, widely differing in length, because were then analyzed with CPIDR and the proposition counts are shown in the accompanying scatterplot. scared **Algorithm initially counts** parts of speech likely to signify propositions (verbs, adjectives, conjunctions, etc.). Results Adjustment rules remove the linking verb was from the count. CPIDR agreed with the group of human raters appreciably better than the raters agreed with each other (r = 0.942, or 0.969 if one outlier is excluded, vs. $r \ge 0.82$ for human vs. human). **Proposition counts of** CPIDR's proposition counts ran about 5% lower, **40 text samples** on average, than those from the human raters (see regression line in scatterplot). This is probably due mainly to a stricter interpretation 200 Г of Turner and Greene (1977). The human raters One speaker who frequent sometimes counted auxiliary and linking verbs. used like as an interjection One text, shown as an outlier on the graph, was greatly overcounted by CPIDR. The explanation Regression line y = 0.945 xis that this speaker frequently used *like* as an (neglecting one outlier) *r* = 0.942 almost meaningless interjection, which the (0.969 if one outlier is human raters correctly skipped. CPIDR, omitted) however, using a tagger trained on the Penn Treebank, treated *like* as a verb or preposition. 100 200 150 The regression line on the graph excludes this outlier. **Counted by human raters**

Propositions:

Tagger output	Pre pro cou
l / pronoun	
called / verb	call
my / determiner	my
because / sub. conj.	bec
l / pronoun	
was / verb	was
scared / adjective	scal



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What to do next

Refine the set of adjustment rules and capture all relevant linguistic generalizations

Package CPIDR as a shareable software package

Test CPIDR on other corpora that have been analyzed into propositions by human raters (collaborators welcome!)

Apply CPIDR to our ongoing studies of language in:

- schizophrenia
- learning disorders
- Alzheimer's disease

Factor propositional idea density into its components (verb density, adjective density, etc.) and determine the neuropsychological relevance of each

References

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