



Using Text Mining to Evaluate Quality of Free-Text Health Goals in People with HIV: Proof-of-Concept

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McGill



the CTN
CIHR Canadian
HIV Trials Network

le Réseau
Réseau canadien
pour les essais VIH des IRSC

HIV is now within the list of chronic diseases

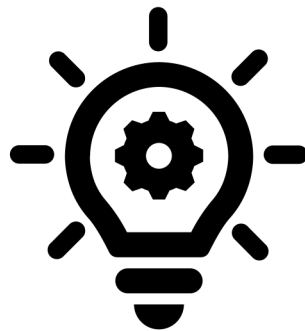
A chronic condition requires day-to-day management by

the person affected

&

GOAL SETTING

- Plays a critical role in day-to-day management of people living with chronic conditions
- Ideal goals are *SMART* (*Specific, Measurable, Actionable, Realistic, & Time-bound*)
- *SMART* goals achieve desired effects on health outcomes and are often set collaboratively
= Clinician + Client



Unsupervised goals lack the structure and format

Measuring goal quality is important

Text mining to the rescue

Objective: To identify whether text mining techniques can be used to discriminate good quality goals

Methods: Two sources of data were tapped. The first was a set of goals set collaboratively (supervised) during a project on health outcomes post-hospitalization. The second source arose from cognitive interviews conducted with people with HIV piloting a goal-setting exercise for a future trial. Half of the goals were uninformed and half informed (i.e., feedback via providing a health profile).

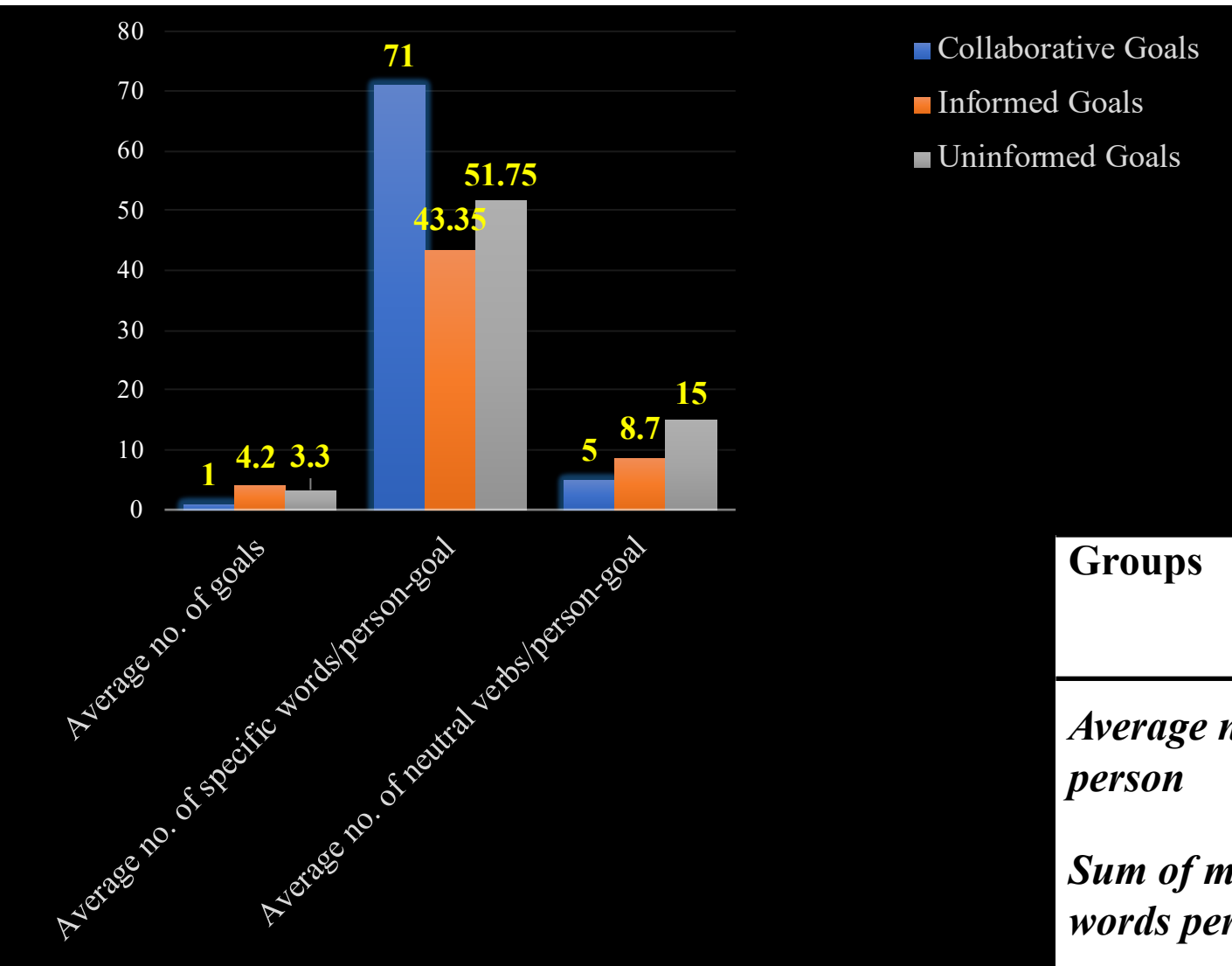


- The main outcome was the extent to which goals were *SMART* by using **specific words and actionable verbs**.
- A measurement framework and an initial lexical (i.e., collection of vocabularies) were developed for the goal evaluation.
- Using text mining techniques, the specific components of each goal (i.e., tokenizing and *pos*-tagging) were extracted and compared to the lexical using regular expression algorithms.
- Text mining offers solution to organize **free text into pre-defined groupings** that can be analyzed to identify patterns in goal-setting quality.
- ***Limitation:** Usability & readability of the profile was being tested via cognitive interviews. Participants did not receive their own health profile. The profile kept changing as comments were received. Despite all, the results are encouraging.



BRAIN HEALTH NOW : Your Personal Brain Health Profile

Participant Number:10-002			
Important Brain Health Areas	Your first visit	Your most recent visit	Optimal
Visit date	October 29, 2013	January 13, 2016	
Cognitive test score	Excellent	Good	Excellent
Your evaluation of your memory	Good	.	Excellent
Able to concentrate	A moderate amount	A moderate amount	Very much to extreme
Negative feelings (blue mood, despair, anxiety, depression)	Quite often	Seldom	Never or seldom
Feeling lonely	Sometimes	Sometimes	Almost never
Time feeling worn out	A little	Some	None to a little
Feeling rested after waking up	Never	Never	Often or always
Your pain rating	None	None	None to mild
Climbing several stairs	Limited a little	Not limited	Not limited
Walking more than a kilometer	Limited a little	Limited a little	Not limited
Vigorous activities	Limited a little	Limited a little	Not limited
Weight to height ratio (BMI)	27	25	Between 19 and 25
Smoking	Smoking	Smoking	Not smoking
Health rating	Very good	Very good	Excellent, very good
Quality of life	Good	Good	Good, very good
<div><div></div> The green boxes indicate areas where you are in the optimal state</div> <div><div></div> The neutral boxes indicate areas where you are below the optimal state but not too far away</div> <div><div></div> The orange boxes indicate areas where you are further away from the optimal state</div> <div>Number of optimal areas at my first visit = 5</div> <div>Number of optimal areas at my most recent visit = 6</div>			



Results: Collaborative goals had more actionable verbs (n/person-goal) than informed goals (n/person-goal), and uninformed goals mainly had neutral verbs.

Rates and rate ratio of goal specificity

Groups	Collaborative (n = 10)	Uninformed (n = 10)
<i>Average number of goals per person</i>	1	3.3
<i>Sum of mean number of specific words per person goal</i>	71	51.75
<i>Rate of specific goals</i>	7.1	1.6
<i>Rate Ratio</i>	4.4 (CI 2.0 – 9.9)	

Comparison of 3 groups

- Specific words and actionable verbs = good quality goal
- Neutral verbs = poor quality goal

Word visualization for “Uninformed Goals”



Word visualization for “Informed Goals”

More words have been taken from the health profile (i.e., use of the health profile to set goals)



Word visualization for “Collaboratively Set Goals”

SMART elements are all present



Conclusion: The process of text mining is ideal for extracting and quantifying specific goal content. With more data, clustering techniques can be applied to shed light on the preoccupations of people with HIV with respect to health outcomes.

* Size of the words represents frequency of their use within the goals.