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## Problem Statement

How and to what extent is it possible to automatically analyze human intentions in textual content?

### Motivation

#### A Novel Perspective:

- existing text analysis techniques provide different perspectives on textual resources, e.g. topics, sentiment or opinion analysis [3]
- Can we analyze textual content from an intentional perspective?

#### The Challenge:

- people rarely state their intentions explicitly in text which makes analyzing human intentions in text a difficult endeavor

#### The Opportunity:

- however, people are quite prolific in writing about the actions and activities they participate in, such as "convert to Christianity", which can be assumed to indirectly contribute to "Achieve Salvation" (taken from [1])
- Can we use these *indicative actions* as a proxy for inferring intentions?

#### Consequences:

- inferring human intentions from textual content could contribute to
  - annotating resources with intentions referenced within them
  - answering "Why" questions in Natural Language Text applications

### Intent Analysis

#### Definition:

- We define Intent Analysis as the problem of automatically identifying a set of adequate intent annotations for textual resources.
- Intent Analysis approximates the unknown function  $f: S \times C^I \rightarrow \{True, False\}$ , where  $C^I = \{c^1, c^2, \dots, c^I, \dots, c^I, \dots, c^I, \dots\}$  is the set of predefined intent categories,  $D$  is a domain of text documents and each document  $d_i$  consists of a sequence of sentences  $S = \{s_1, s_2, \dots, s_{|S|}\}$ .
- yields an *intent profile vector*  $ip \in R^n$ , when applied to a text document, i.e.  $f(d)$

#### Demarcation:

- traditional analysis techniques such as topic, sentiment or opinion analysis [3] aim to approximate a present (e.g. an emotional) state
- in contrast, Intent Analysis deals with a different temporal focus, i.e. a *future state of affairs*, illustrating the difference between **Content vs. Intent**

### Approach: Automatically Generating Intent Profiles

1.) **Enriching a Taxonomy of Human Goals:** We employed the social-psychological theoretical framework [1] that organizes high-level goals of people into 135 categories of human intent including "A good marriage", "Getting an education" and "Taking care of family". We compiled a set of descriptive phrases for each category. To give an example: Descriptive phrases for the category "Achieve Salvation" included "to reach spiritual enlightenment" or "to get into heaven".

2.) **Constructing the Knowledge Base:** We attempted to acquire *indicative actions* by searching for sentences on the web (cf. [2]). We constructed a series of query strings by concatenating each descriptive phrase with causal relation phrases, e.g. "in order to". Then, exact phrase searches were issued to the web using Yahoo! BOSS. We identified ~169.000 sentences that contained the query phrases along with indicative actions. These sentences were stored in our knowledge base, an Apache Lucene index.

3.) **Matching Sentences to Intent Categories:** We first segment the document into a set of sentences. Each sentence in the document is issued as a query to the knowledge base to identify the most similar sentence in our knowledge base.

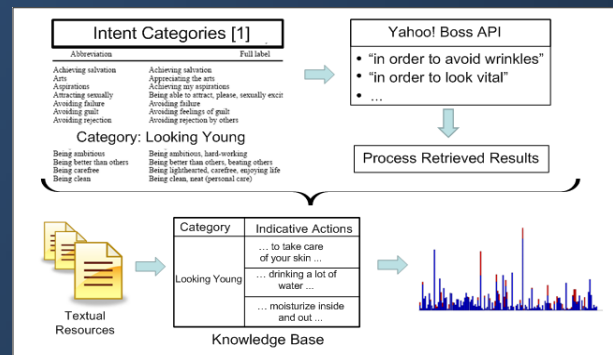


Figure 1: This figure visualizes the process of automatically analyzing textual resources and generating intent profiles from textual resources.

### Results & Evaluation: Automatically Analyzing B. Obama's and J. McCain's US Presidential Election Speeches

We applied our approach to the textual resources of 44 transcripts of political speeches given by the two American presidential candidates in 2008. Figure 2 provides a visual comparison between the two candidates intent profiles. To evaluate the quality of automatic intent annotations, we compared notations produced in a human subject study.

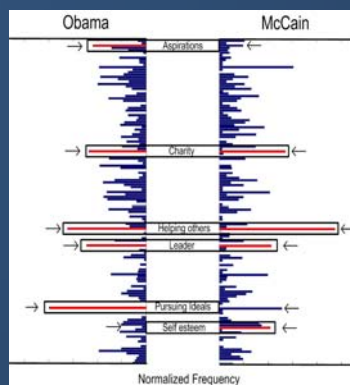


Figure 2: Intent profiles of Obama and McCain, generated from and averaged over 44 speeches (April and June 2008). It illustrates similarities and differences between the two candidates that can be easily recognized (intentional summary).

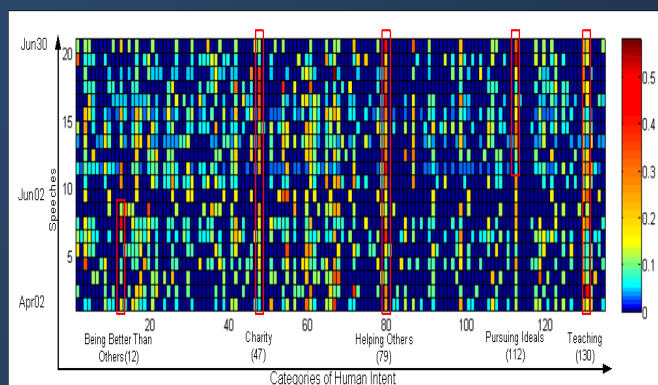


Figure 3: This visualization allows a comparison of intent profiles over time. The intent profiles were generated from 21 speeches given by Barack Obama in April and June 2008. It shows the temporal evolution of intent in the political speeches. Categories such as "Charity" and "Helping Others" are prominent over the whole period. Peaks such as the category "Pursuing Ideals" can be easily detected.

Automatic Annotation	Rank	Manual Annotation	Rank
Helping others	0.2368	Avoiding failure	0.0598
Being better than others	0.1513	Aspirations	0.0949
Charity	0.1335	Standing up for beliefs	0.0873
Pursuing Ideals	0.1278	Helping others	0.0863
Leader	0.1058	Being respected	0.0852
Self esteem	0.1039	Pursuing Ideals	0.0846
Ethical	0.103	Being recognized	0.0843
Money	0.099	Persuading others	0.0883
Being socially attractive	0.0919	Being responsible	0.0862
Seeking justice	0.0862	Overcoming failure	0.0819
Seeking fairness	0.0811	New ideas	0.0809
Being intelligent	0.0805	Own guidelines	0.0777
Easy life	0.0773	Leader	0.0766
Belonging	0.0747	Support from others	0.0766
Career	0.0738	Being better than others	0.0793
Peace of mind	0.0673	Control over others	0.0781
Being honest	0.0653	Teaching	0.072
Teaching	0.0651	Others' trust	0.071
Feeling safe	0.0643	Seeking fairness	0.071
Being respected	0.0636	Being honest	0.076
Being creative	0.059	Seeking justice	0.075
Good parent	0.0567	Freedom of choice	0.0728
Personal growth	0.0543	Career	0.0728
Content with myself	0.0529	Seeking equality	0.071
Being responsible	0.0525	Taking care of family	0.071

Figure 4: Comparison of the top 25 Intent Categories (automatically and manually generated) for 23 speeches of J. McCain. The results illustrate that 11 categories (44%) are shared between automatic and human annotations.

### Contributions

- This work
- adds a novel dimension to the repertoire of textual data analysis techniques
- expands the knowledge that can be inferred from textual resources
- demonstrates the possibility of automated intent analysis
- indicates the potentials of Intent Analysis as a quick, visual evaluation of text from an intentional perspective

### Further Questions

- To what extent is intent analysis applicable to other text corpora such as weblogs?
- To what extent does our approach contribute to the notion of social media monitoring?

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[1] Chulef, A. S., Read, S. J. and Walsh, D. A.: A Hierarchical Taxonomy of Human Goals. In *Motivation and Emotion*, 25 (3), 191–232, 2001.  
[2] Gimino, P. and Szab, S.: Learning by googling. *SIGMOD Explor. Newsl.* 6(2), 74–83, 2004.  
[3] Pang, B. and Lee, L. Opinion mining and sentiment analysis. In *Foundations and Trends® in Information Retrieval* 2(1-2), 1–135, 2008.