Text Preparation through Extended Tokenization

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Overview

- Introduction
- Token Concepts
- Extended Tokenization
- Token Typing
- JavaTok
- Conclusion

Introduction

Tokenization

- o first step of NL text preparation
- stream of characters \rightarrow stream of tokens (processing units)
- Supports any further NLP task
 - Tagging, Named Entity Recognition, Parsing, etc.



- Task taken for granted
 - already solved problem
 - theoretically uninteresting
 - without large impact (e.g., Information Retrieval)

Introduction

Standard tokenization algorithm

- 1) split strings separated by blanks / linefeeds
- 2) split all punctuation marks ('.', '!', '?') ending those strings

Difficulties of the standard algorithm

- o unclear token borders: doesn't
- o sentence borders: der 15. Platz, semicolons
- o abbreviations: e.g., etc.

Assumptions

- o language independent
- o domain independent
- application independent

Token Concepts

Single-tokens

• Strings without non-printable or delimiting characters

• Examples:

- single words: car, information, Sidney
- numbers: 12345, 12.43, 8, 45
- internet addresses: http://www.google.com
- Multi-tokens
 - Strings through interpretation (may contain delimiters)
 - Examples:
 - composite nouns: traffic jam, information retrieval
 - special formats: +43 463 2700-3511, ISDN-12 34567 / 89
 - named entities: United States of America
 - formulas: \$x = x+1\$

Extended Tokenization

- Do as much as possible on STRING level, but not more!
- Extended Tokenization process
 - 1) identify single-tokens (standard tokenization)
 - 2) type single-tokens
 - 3) identify sentence end markers
 - 4) reinterpret single-token types
 - 5) merge and split tokens recursively (multi-tokens)
 - 6) reinterpret any token type

Extended Tokenization

Incorporate many kinds of linguistic knowledge like

- semantically motivated string patterns
 - e.g., phone numbers, serial codes, dates, URLs
- o dictionaries
 - e.g., abbreviations, names
- o morphosyntactic and sentence related rules
 - e.g., derivation (cold coldness), composition (scarface), capitalized term must start a new sentence

Resources

- o language dependent
- o domain specific
- o application oriented



Pre-linguistic classification process

3 step typing process

- 1) type single-tokens (basic token types)
- reinterpret single-token types (user-defined token types)
- reinterpret token types (user-defined token types)

Token Typing: Basic Types

Assigned straight-forward

- Basic token types (4)
 - o alphabetics: test, Test, TEST, TesT
 - o numerics: 123, 12.3, 1, 23, 12:34
 - punctuation marks
 - sentence end marker
 - sentence-internal marks like comma
 - pair wise markers like brackets and quotes
 - o mixtures
 - ending with sentence end marker
 - starting/ending with hyphen
 - containing slashes / hyphens
 - containing numbers
 - others

Token Typing: User-defined Types

User-defined Token Types

- expressed through strings
- o identified by rules and minimal dictionary knowledge

Includes

- o domain knowledge
 - e.g., knowledge about data warehouses
- o gazetteer knowledge
 - e.g., country names, organization names
- o expert knowledge
 - e.g., medicine
- o pure linguistic knowledge
 - e.g., morphological and syntactical rules

Token Typing: User-defined Types

Examples

- o abbreviations
- o acronyms
- o dates and times
- o phone numbers
- email addresses
- sequences of capitalized single-tokens (NE candidates)
- o stopwords
- o etc.

Prototype

- fully implemented in Java
- o part of the NLP toolset actually developed
- o online demo available at: http://nlp.ifit.uni-klu.ac.at/NLP/

Features

- free configuration and adaptation (UTF-16)
- o completely rule-based with dictionary support
- o enables user-defined token type definition
- string replacements (abbreviation resolution, zero elimination, thesaurus, ...)
- o pre-tagging functionality (based on token types)
- o multiple output formats (TXT, HTML, XML)

Rules

- o applied on token strings, token types, or both
- support RegEx matching / substitution
- access arbitrary long sequence of tokens

Examples

- o suffix identification of well-known endings (e.g., -ly, -ness).
- identification and reconcatenation of hyphenated words
- o sentence border disambiguation
- multi-token identification
- o special character treatment, e.g., & % \$ § ° ` ∖ /

Example text output for

The Red Cross is aka. RK.

Output The Red Cross is aka. RK . The Red Cross is also known as RK . The (Red Cross)/INST is aka. RK . The (Red Cross)/INST is (also known as)/ABBR RK . The/T _{a2} Red/T _{a2} Cross/T _{a2} is/T _{a1} aka./ABBR RK/T _{a3} ./T _{p1} The/T _{a2} Red/T _{a2} Cross/T _{a2} is/T _{a1} also/T _{a1} known/T _{a1} as/T _{a1} RK/T _{a3} ./T _{p1}			
The Red Cross is also known as RK . The (Red Cross)/INST is aka. RK . The (Red Cross)/INST is (also known as)/ABBR RK . The/T _{a2} Red/T _{a2} Cross/T _{a2} is/T _{a1} aka./ABBR RK/T _{a3} ./T _{p1}	s	М	R
The (Red Cross)/INST is aka. RK . The (Red Cross)/INST is (also known as)/ABBR RK . The/T _{a2} Red/T _{a2} Cross/T _{a2} is/T _{a1} aka./ABBR RK/T _{a3} ./T _{p1}			
The (Red Cross)/INST is (also known as)/ABBR RK . The/ T_{a2} Red/ T_{a2} Cross/ T_{a2} is/ T_{a1} aka./ABBR RK/ T_{a3} ./ T_{p1}			х
The/T _{a2} Red/T _{a2} Cross/T _{a2} is/T _{a1} aka./ABBR RK/T _{a3} ./T _{p1}		х	
		х	х
The T_{a2} Red T_{a2} Cross T_{a2} is T_{a1} also T_{a1} known T_{a1} as T_{a1} RK T_{a3} . T_{p1}	х		
	х		х
The T_{a2} (Red T_{a2} Cross T_{a2}) / INST is T_{a1} aka. / ABBR RK T_{a3} . T_{p1}	Х	х	
$\label{eq:tailor} The/T_{a2} ~ (\text{Red}/T_{a2} ~ \text{Cross}/T_{a2}) / \text{INST} ~ \text{is}/T_{a1} ~ (also/T_{a1} ~ known/T_{a1} ~ as/T_{a1}) / \text{ABBR} ~ \text{RK}/T_{a3} ~ . / T_{p1}$	Х	х	х

S = single-token typing, M = multi-token typing, R = replacement of strings

Preliminary results

- o improvements of tagging outputs for
 - Stanford ME tagger
 - openNLP Tools ME tagger
 - QTag
- Corpus-based training (rule generation)
 - o INEX (INitiative for the Evaluation of XML retrieval) collection

Further steps

- large scale evaluation
- o compare results to others

Conclusion

- Proper tokenization is crucial for any further NLP task
- Relies on the token definition
- Supported by rule-based token typing
- Online implementation JavaTok
 - o http://nlp.ifit.uni-klu.ac.at/NLP/