Computational Linguistics in the Industry

Authoring Support with acrolinx IQ ™



overview

- > acrolinx the company
- > production of technical documents
- > NLP for
 - >spelling and terminology
 - ≻grammar
 - ≻style
 - consistent phrasing





acrolinx - the company

- software for information quality assurance
- spin-off from German Research Center for Artificial Intelligence (DFKI), Saarbrücken
- technology under development since 1997 (since 2002 as acrolinx))
- headquarter in Berlin, about 40 employees
 users in 25 countries, checking millions of
- words a month





some of our customers

Software	Life Sciences	Communicatio ns	Industrial	Technology
Adobe	Dräger	AlcatelLucent	DAF	Bosch
Autodesk	GE	Cisco	HOMAG	Embraer
CA	Medtronic	Huawei	John Deere	KonicaMinol ta
EMC	Siemens	Motorola	MAN	Philips
IBM		SonyEricsson	SEW Eurodrive	
SAS Institute			Siemens	
Symantec			Leica GeoSystems	

production of technical documents

- correctness
- understandability
- readability
- translatability
- consistence
- Iess ambiguity
- corporate wording

- spelling
- grammar
- style
- terminology

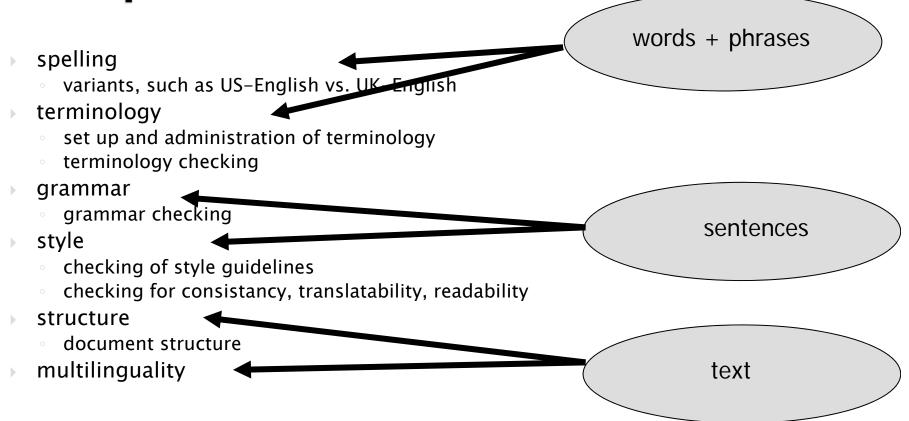
QA is a Cost Factor

Translation costsSupport costs





text production







spelling

words are defined in a dictionary anything not in the dictionary is an error high recall, low precision (depending on the domain) errors are defined
 unknown words that are not defined as errors are term candidates

- based on words and rules
- consider terminology

 high precision, recall is dependent on data work

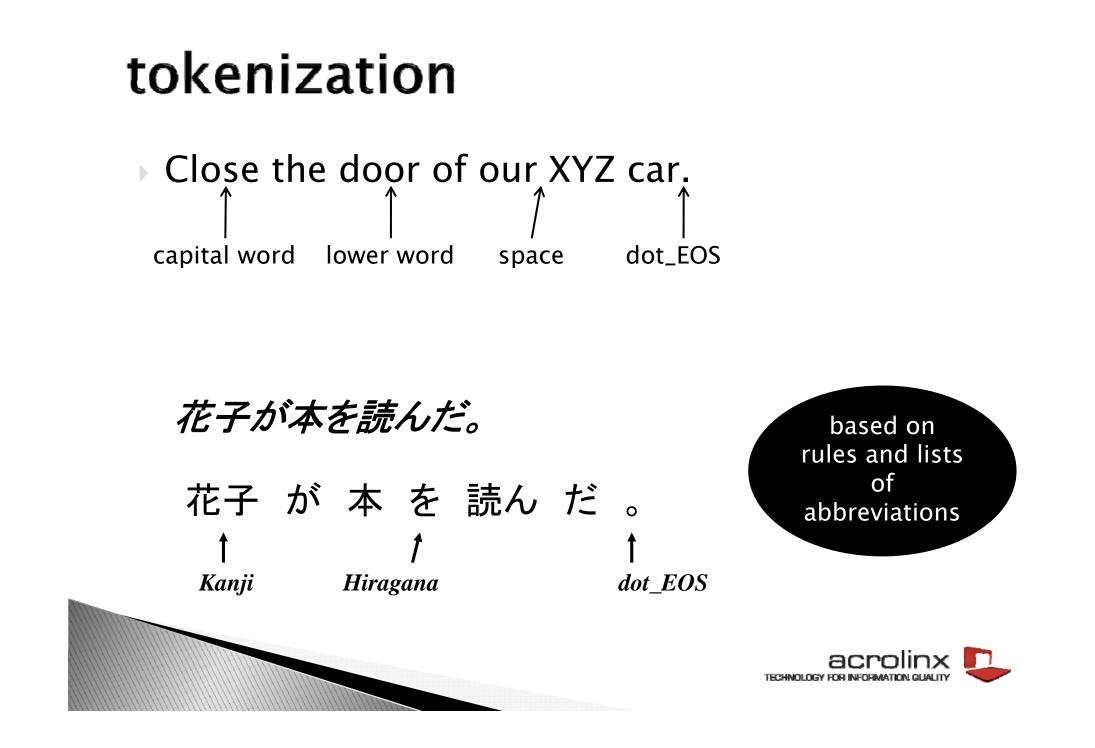
language analysis	error analysis

NLP for words and phrases

- tokenization
- POS-tagging
- morphology
- dictionary
- error dictionary

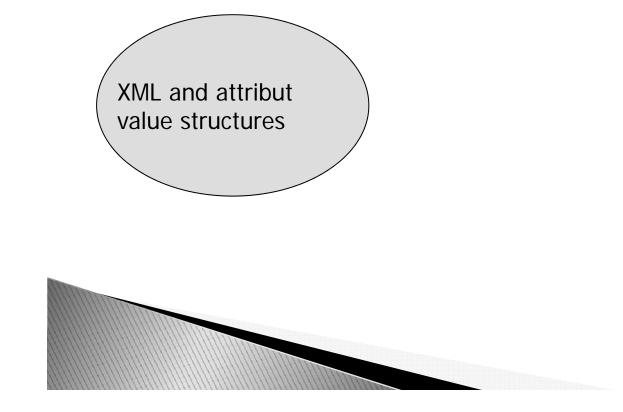


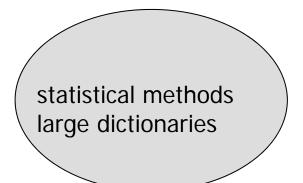




POS tagging, such as:

Close the door of our XYZ car.V DET N PREP PRON NE N







morphology

Close the door of our XYZ car.

Lemma: close Tense: present_imp Person: third Number: singular

Lemma: car Number: singular Case: nominative_accusative

based on dictionaries, rules for inflection and derivation



terminology: Why work on terminology?

- Consistency!
- ideally: 1 term = 1 meaning = 1 translation
- less ambiguity, better comprehension, translatability, etc.
- multilingual consistency
- corporate wording
- Iower costs (translation but also support)



the reality ...

- When analyzing terminology in documents, we find many variants that are used at the same time:
 - web server web-server
 - upload protection upload-protection
 - timeout time out
 - Reset ReSet
 - sub station sub-station





how to get consistent terminology?

- author/company defines term banks
- list of deprecated terms deprecated term: vehicle approved term: car

 ▶ list of approved terms
 → identification of so-called "variants" approved term: SWASSNet User deprecated term: SWASSNet user, SWASS-Net User



term variants

- orthographic variants
 - hyphen, blank, case: term bank, termbank
- semi-orthographic variants
 - number : 6-digit, six-digit
 - trademark : acrolinx IQ[™], acrolinx IQ
- syntactic variants
 - preposition: oil level, level of oil
 - gerund/noun : call center, calling center
- synonyms

"classical" : vehicle, car

 language-specific variants (e.g. Fugenelemente DE, Katakana JA)



terminology and spelling

in terminology: SpeicherKarte

- Erste Schritte¶
- Installieren der SIM-Karte, des Akkus und der SpeicherKarte¶
- Installieren der SIM-Karte, des Akkus und des SpeicherModuls





terminology and spelling

term: MMC-Speicherkarten (deprecated), suggested: PC-Speicherkarten

Verwenden Sie nur kompatible MMC-Speicherkrten (MultiMediacard) mit diesem Gerät.

Help

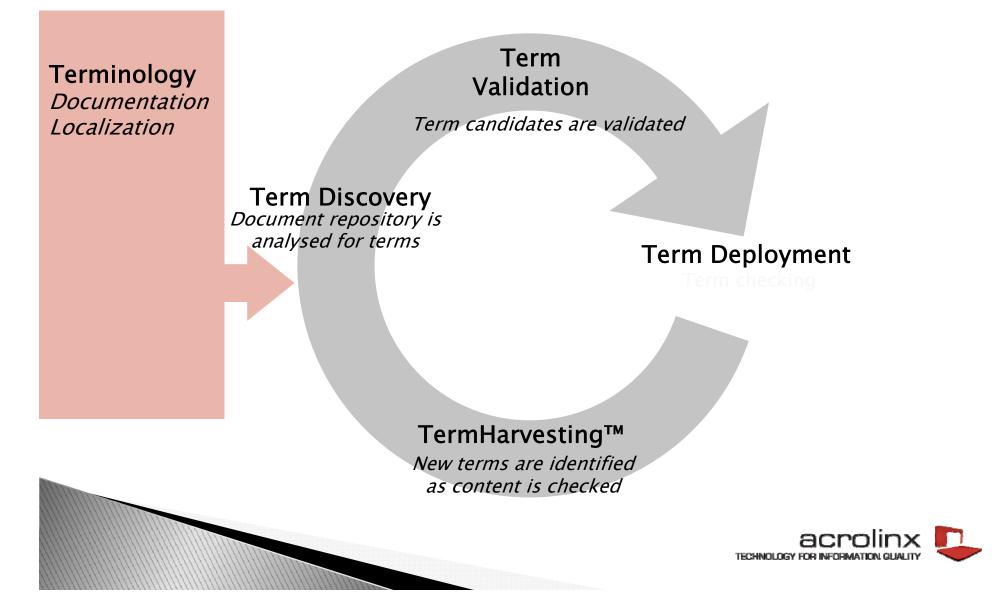
Andere Speicherkarten wie SD-Karten mit diesem Gerät nicht kompatibel. Durc Speicherkarte kann sowohl die Karte als können die auf auf der nicht kompatibler

- Bevor Sie den Akku herausnemen, m vom Ladegerät trennen.¶
- Wenden Sie das Gerät, so daß die Rü Entriegelungstaste und schieben Sie

Replace with :	steckplatz und sind kompatiblen rden. <mark>Außerdem</mark> schädigt werden. ¶		
PC-Speicherkarten			
Edit Flag			
Ignore Flag	·		
Step-through Mode	ausschalten und es		
Previous Flag	en Sie auf die		
Next Flag			
Standards and Guidelines			



terminology lifecycle management



NLP for terminology

- NLP methods for term extraction
 - corpus analysis (morphology, POS, NER)
 - information extraction (potential product names)
 - ontologies (e.g. semantic groups)
- NLP methods for setting up a term database
 - morphology (finding the lemma)
 - POS
- NLP methods for term checking
 - variants
 - similar words
 - inflection





approaches to grammar checking

definition of correct grammar

- e.g. HPSG, LFG, chunkgrammar, statistical grammars
 anything that's not analyzable
- must be a grammar error preconditions:
 - grammar with large coverage
 - giant dictionaries
 - robust, but not too robust parsing
 - efficient parsing methods
- high recall, low precision

grammar errors are implemented

- preconditions:
 - work with error corpora
 - error grammar with a high number of error types
 - "deepness" of analysis varies with the type of error to be described
- high precision, recall is based on the number of rules

descriptive grammar

error grammar

grammar rules, for example

subject verb agreement:

- Check if instructions are programmed in such a way that a scan never finish.
- When the operations is completed, the return to home completes.

a an distinction:

- a isolating transformer
- an program

wrong verb form:

- it cannot communicates with them
- IP can be automatically get



example grammar rule

```
write_words_together
```

```
 @can ::= [ TOK "^(can)$"
 MORPH.READING.MCAT "^Verb$" ]:
```

```
• The application can not start.
```

```
 The application can tomorrow not start.
```

```
 TRIGGER(80) == @can^1 [@adv]* 'not'^2

         -> ($can, $not)
         -> { mark: $can, $not;
         suggest: $can -> ", $not -> 'cannot';
         }
```

 Branch circuits can not only minimize system damage but can interrupt the flow of fault current

```
 NEG_EV(40) == $can 'not' 'only' @verbInf []* 'but';
```



style - controlled language

- controlled languages
 - AECMA now: AeroSpace and Defence Industries Association of Europe (ASD) ASD-STE100 (simplified English)
 - Caterpillar Technical English (CTE)
- disadvantage:
 - very restrictive! Prescriptive rules define allowed structures and allowed vocabulary → all other structures and words as disallowed
 - low acceptance of user





style - error definition

- rules define errors (just as grammar rules do)
- rules are defined by user / author
- acceptance is much higher





style

- style guidelines can be different for different usages
 - text type (e.g., press release technical documentation)
 - domain (e.g., software machines)
 - readers (e.g., end users service personnel)
 - authors (e.g., Germans tend to write long sentences)





style rule examples: best practise

- avoid_latin_expressions
- avoid_modal_verbs
- avoid_passive
- avoid_split_infinitives
- avoid_subjunctive
- use_serial_comma
- use_comma_after_introductory_phrase
- spell_out_numerals





style rule examples: company

- use_units_consistently
- abbreviate_currency
- COMPANY_trademark
- do_not_refer_to_COMPANY_intranet
- add_tag_to_UI_string
- avoid_trademark_as_noun
- avoid_articles_in_title



style rule examples MT (preediting)

- avoid_nested_sentences
- avoid_ing_words
- •keep_two_verb_parts_together
- avoid_parenthetical_expressions
 - dependent of MT system and language pair



style rule suggestions

- replacement of words or phrases
- replacement using the correct writing with uppercase or lowercase
- replacement of words using the correct inflection
- generation of whole sentences (e.g. passive active) requires semantic analysis and generation and is therefore not (yet) possible





example of style rule

avoid_future

/* Example: "... It will be necessary ..." */

TRIGGER (80) == @will^1 [-@comma]* @verbInf^2
 ->(\$will, \$verbInf)
 -> { mark : \$will, \$verbInf;}

/* Example: "... The router services will be offered in the future ..."*/

NEG_EV(40) == \$will []* @next @time;



consistent phrasing: why?

- Use the same phrase for the same meaning.
- Examples:
 - Congratulations on acquiring your new wearable digital audio player
 - Congratulations, you have acquired your new wearable digital audio player!
 - Dear Customer, congratulations on purchasing the new wearable digital audio player!
- Using the same phrase makes the documents more readable and helps to save translation costs.



the reality...

- End date must be equal to or later than the start date.
- End Date must be greater than or equal to Start Date.
- End Date must be greater than Start Date.
- End Date must be later than Start Date.
- End date should be greater than start date.
- End Date cannot be before the Start Date.
- Please enter an end date that is later than the start date.
- Please enter an End Date that is later than or the same as the Start Date.
- Please enter a start date that is before the end date.
- Start date must be before end date!
- The end date must be later than or the same as the start date.
- The start date cannot be later than the end date.
- The start date must be on or before the end date.
- The Start Date cannot be after the End Date.
- The end date cannot be before the start date.
- The actual end date must be on or after the actual start date.
- The start date must be prior to the end date.
- The ending date must be later than or the same as the beginning date.
- Your end date must be after your start date.
- You cannot enter an "End Date" that is before your "Start Date."
- Your start date must be before your end date.
- You entered a start date later than the end date.



Intelligent Reuse

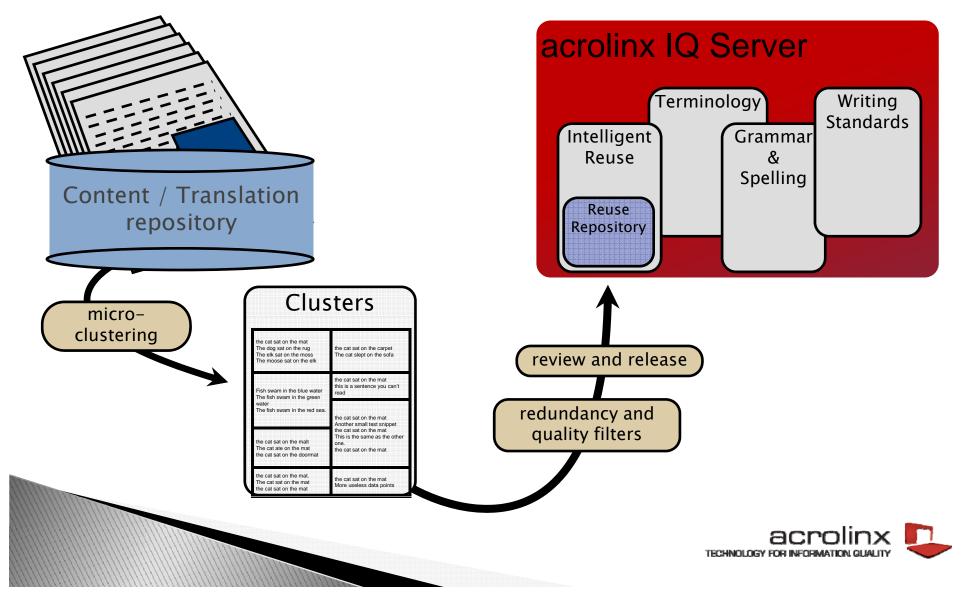
- analysis of text documents with NLP, such as ontologies, morphology, sentence similarity
 - selection of standard sentences
 - automatic selection with respect to grammar, style, terminology
 - human validation
 - suggestions for similar sentences in new texts





Intelligent Reuse™

Building your Reuse Repository



NLP components in acrolinx

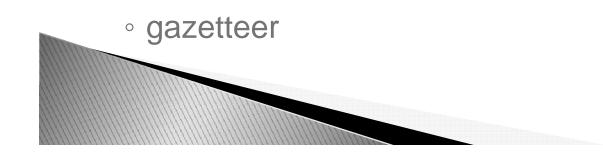
components for analysis

• tokenizer (sentences and words)

morphology, decomposition

• POS tagger

• word guesser





NLP components in acrolinx

- rule formalism is based on language analysis results
 - spelling
 - grammar
 - style
 - term variants
 - term extraction





