Natural Language Processing (overview)

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Lecture outline

- What is NLP?
- Applications & Approaches
- Resources used in NLP
- NLP subtasks; Ambiguity
- Evaluation in NLP
- Precision grammar engineering
- NLP around UW

What is NLP?

Processing language by computers
 Distinct from speech processing
 Not necessarily linguistically motivated

Applications (1/3)

- Linguistic research
- Grammar checking/spell checking
- Computer assisted language learning (CALL)
- Assistive & augmentative communication (AAC)

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Applications (2/3)

• Machine translation, machine assisted translation

• Information retrieval

• Information extraction

-- Monolingual & multilingual

Applications (3/3)

- HCI
 - Natural language database access
 - UI navigation
 - Automated customer service
 - Games
- Other?

Approaches

Knowledge engineering
Machine learning
Hybrid

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Resources

- *Oictionaries (monolingual, bilingual) Corpora*
- Annotated corpora
 - Tagged corpora (POS, word sense,...)
 - Treebanks
 - Aligned bilingual/multilingual corpora

Useful for...

- Supervised learning
- Gold standard/evaluation
- Unsupervised/semi-supervised learning of the next layer of linguistic structure
- Linguistic hypothesis testing

Sources of Resources

- LDC: Linguistic Data Consortium
- ELDA: Evaluations and Language resources Distribution Agency
- Rosetta: All Languages Archive

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NLP subtasks (1/3)

Language identification
Part of speech tagging
Word sense disambiguation
Named entity recognition
NP/other phrase detection

NLP subtasks (2/3)

- Stemming/morphological analysis
- Segmentation (documents to sentences, sentences to words)
- Sentence, phrase, word alignment (of bitext)

NLP subtasks (3/3)

- Parsing (string to tree; string to semantics)
- Generation (semantics to string)
- Reference resolution
- Speech act recognition
- Dialogue planning
- Others?

Ambiguity

- Natural language wasn't designed to be processed by computers.
- Ambiguity (local and global) at every level of structure
- Potentially want to return multiple analyses
- ... while also being able to rank them

Ambiguity examples

• Word boundary: Dungeon of Spit • Part of speech: read, record, talk • Morphological analysis: kayaking, singing, sing, anything walks, unwrappable

More ambiguity examples

• Syntax: Kim is our local unicode expert. Have that report on my desk by Friday. • Semantics: Every cat chased some dog. • Speech act: Can you pass the salt?

Still more ambiguity examples

 Reference resolution: The police denied the protesters a permit because they feared/advocated violence.

 String realization: Kim gave the dog a bone. Kim gave a bone to the dog.
 Addressee recognition

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Evaluation

- Requires:
 - Test set with gold standard answers for comparison
 - Metric(s) of comparison
 - Baseline strategy to compare against
- All three of these can be non-obvious in NLP

Evaluation

- Validation: Does my system behave the way I think it behaves?
- Regression testing: What did I break today?
- Experimental results: How does my system compare to other systems?

Humans are expensive

• Evaluation processes should be automated wherever possible:

- Speed
- Cost

• Integration into development cycle

Easy case: POS tagging

- Gold standard: A corpus with POS tags annotated (by humans)
- Evaluation metric: Precision (number of correct tags/total tags)
- Baseline: Random assignment of possible tags for each lexical item
- Wrinkle: Count performace on unambiguous items?

Harder case: Parsing

- How to create a gold standard?
- Sources of variation:
 - Genuine structural ambiguity (usually, but not always, resolved in context)
 - Different styles of representation/ different linguistic theories



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Harder case: Parsing

- In practice, the most common gold standard is the Penn Treebank
- 1 million words of hand-parsed Wall
 Street Journal text + 1 million words of
 hand-parsed Brown corpus
- More or less internally consistent; not consistent with any particular linguistic theory

Harder case: Parsing

- Evaluation metric?
 - How many sentences got exactly the gold standard tree
 - A more sophisticated solution is PARSEVAL (there are others)

PARSEVAL

• Labeled precision:

correct constituents in candidate parse consituents in candidate parse

• Labeled recall:

correct constituents in candidate parse consituents in gold standard parse

• Crossing brackets:

(A (B C)) v ((A B) C)

Harder case: Parsing

- What would be a sensible baseline?
- Randomly choosing among all possible structures assigned by the grammar
- Comparison to other existing systems

Even harder case: MT

- (NB: Human evaluation is particularly expensive in this case.)
- What should be the gold standard?
- Are all things that differ from the gold standard necessarily wrong?
- More so than with parsing?

MT and BLEU

 Papineni et al 2002: Bleu: a Method for Automatic Evaluation of Machine Translation

 A good translation will have a distribution of n-grams similar to other good translations

BLEU

• Modified n-gram precision

$$\frac{\sum_{C \in candidates} \sum_{n-gram \in C} Count_{clip}(n-gram)}{\sum_{C \in candidates} \sum_{n-gram' \in C'} Count_{clip}(n-gram')}$$

• Geometric mean of n-gram precisions for different N, plus a brevity penalty

Evaluation in NLP summary

- It's always important
- The nature of the tasks makes it often hard to define a gold standard and evaluation metric
- Gold standards can also be expensive

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Natural language syntax & semantics

- Constituent structure
- Mapping of linear string to predicateargument structure (word order, case, agreement)
- *Long distance dependencies*What did Kim think Pat said Chris saw? *Idioms, collocations*

Formal/'Generative' Grammars

- Characterize a set of strings (phrases and sentences)
- These strings should correspond to those that native speakers find acceptable
- Assign one or more syntactic structures to each string
- Assign one or more semantic structures to each string

Formal/'Generative' Grammars

• No complete generative grammar has ever been written for any language

Precision Computational Grammars

- Knowledge engineering of formal grammars, for:
- Parsing: assigning syntactic structure and semantic representation to strings
- Generation: assigning surface strings to semantic representations

Hurdles

- Efficient processing
- Ambiguity resolution
- Domain portability

(Oepen et al 2002)

(Baldridge & Osborn 2003, Toutanova et al 2005, Riezler et al 2002)

(Baldwin et al 2005)

- Lexical acquisition (Baldwin & Bond 2003, Baldwin 2005)
- Extragrammatical/ungrammatical input (Baldwin et al 2005)
- Scaling to many languages

The Grammar Matrix: Overview

MotivationHPSG

• Semantic representations

• Cross-linguistic core

• Modules

Matrix: Motivation

- English Resource Grammar:
 - 140,000 lines of code (25,000 exclusive of lexicon)
 - ~3000 types
 - 16+ person-years of effort
- Much of that is useful in other languages
- Reduces the cost of developing new grammars

Matrix: Motivation

• Hypothesis testing (monolingual and cross-linguistic)

• Interdependencies between analyses

 Adequacy of analyses for naturally occurring text

Matrix: Motivation

- Promote consistent semantic representations
 - Reuse downstream technology in NLU applications while changing languages
 - Transfer-based (symbolic or stochastic MT)

HPSG

- Head-Driven Phrase Structure Grammar (Pollard & Sag 1994)
- Mildly-context sensitive (Joshi et al 1991)
- Typed feature-structures
- Declarative, order-independent, constraint-based formalism

An HPSG consists of

- A collection of feature-structure descriptions for phrase structure rules and lexical entries
- Organized into a type hierarchy, with supertypes encoding appropriate features and constraints inherited by subtypes
- All rules and entries contain both syntactic and semantic information

An HPSG is used

- By a parser to assign structures and semantic representations to strings
- By a generator to assign structures and strings to semantic representations
- Rules, entries, and structures are DAGs, with type name labeling the nodes
- Constraints on rules and entries are combined via unification

Example rule type

head-subj-phrase: binary-headed-phrase & head-compositional SUBJ COMPS 1 (2)
1 SUBJ HEAD-DTR COMPS N-HEAD-DTR 2

Example rule type

head-final: binary-headed-phrase & HEAD-DTR 1 NON-HEAD-DTR 2 ARGS (2,1)

subj-head: head-subj-phrase & head-final

Example parse



Semantic Representations

- Not going for an interlingua
- Not representing connection to world knowledge
- Not representing lexical semantics (the meaning of life is life')
- Making explicit the relationships among parts of a sentence

Semantic Representations

- Kim gave a book to Sandy
- give(e,x,y,z), name(x, 'Kim'), book(y), name(z, 'Sandy'), past(e)

Semantic Representations

- Sandy was given a book by Kim.
- A book was given to Sandy by Kim.
- Kim continues to give books to Sandy.
- This is the book that Kim gave Sandy.
- Which book did Kim give Sandy?
- Which book do people often seem to forget that Pat knew Kim gave to Sandy?
- This book was difficult for Kim to give to Sandy.

Semantic representations

- Languages may still differ:
 Lexical predicates
 Japanese: kore, sore, are
 - Grammaticized tense/aspect, discourse status
 - Ways of saying
 make a wish, center divider

Matrix Architecture

- Cross-linguistic core encoding language universals
- Set of mutually-compatible 'modules' encoding recurring, but non-universal patterns
- Rapid prototyping of precision grammars
 Ongoing development through Ling 567

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- Professional MA in Computational Linguistics http://www.compling.washington.edu
- Computational Linguistics Lab http://depts.washington.edu/uwcl
- Turing Center http://turing.cs.washington.edu
- SSLI Lab http://ssli.ee.washington.edu
- MS/UW Symposium in Computational Linguistics http://depts.washington.edu/uwcl/msuw/symposium.html
- iSchool, Med School, ...