

Feature-Based Grammar

Ling571

Deep Processing Techniques for NLP

February 3, 2016

Features in CFGs: Agreement

- Goal:
 - Support agreement of NP/VP, Det Nominal
- Approach:
 - Augment CFG rules with features
 - Employ head features
 - Each phrase: VP, NP has head
 - Head: child that provides features to phrase
 - Associates grammatical role with word
 - VP – V; NP – Nom, etc

Simple Feature Grammars

- $S \rightarrow NP[NUM=?n] VP[NUM=?n]$
- $NP[NUM=?n] \rightarrow N[NUM=?n]$
- $NP[NUM=?n] \rightarrow PropN[NUM=?n]$
- $NP[NUM=?n] \rightarrow Det[NUM=?n] N[NUM=?n]$
- $Det[NUM=sg] \rightarrow \text{'this'} \mid \text{'every'}$
- $Det[NUM=pl] \rightarrow \text{'these'} \mid \text{'all'}$
- $N[NUM=sg] \rightarrow \text{'dog'} \mid \text{'girl'} \mid \text{'car'} \mid \text{'child'}$
- $N[NUM=pl] \rightarrow \text{'dogs'} \mid \text{'girls'} \mid \text{'cars'} \mid \text{'children'}$

Parsing with Features

- `>>> cp = load_parser('grammars/book_grammars/feat0.fcfg')`
- `>>> for tree in cp.parse(tokens):`
 - `... print(tree)`
- `(S[] (NP[NUM='sg']`
 - `(PropN[NUM='sg'] Kim))`
 - `(VP[NUM='sg', TENSE='pres']`
 - `(TV[NUM='sg', TENSE='pres'] likes)`
 - `(NP[NUM='pl'] (N[NUM='pl'] children))))`

Feature Applications

- Subcategorization:
 - Verb-Argument constraints
 - Number, type, characteristics of args (e.g. animate)
 - Also adjectives, nouns
- Long distance dependencies
 - E.g. filler-gap relations in wh-questions, rel

Morphosyntactic Features

- Grammatical feature that influences morphological or syntactic behavior
 - English:
 - Number:
 - Dog, dogs
 - Person:
 - Am; are; is
 - Case:
 - I – me; he – him; etc
 - Countability:

Semantic Features

- Grammatical features that influence semantic (meaning) behavior of associated units
- E.g.:
 - ?The rocks slept.
- Many proposed:
 - Animacy: +/-
 - Natural gender: masculine, feminine, neuter
 - Human: +/-
 - Adult: +/-
 - Liquid: +/-

Aspect (J&M 17.4.2)

- The climber hiked for six hours.
 - The climber hiked on Saturday.
 - The climber reached the summit on Saturday.
 - *The climber reached the summit for six hours.
-
- Contrast:
 - Achievement (in an instant) vs activity (for a time)

Unification and the Earley Parser

- Employ constraints to restrict addition to chart
- Actually pretty straightforward
 - Augment rules with feature structure
 - Augment state (chart entries) with DAG
 - Prediction adds DAG from rule
 - Completion applies unification (on copies)
 - Adds entry only if current DAG is NOT subsumed

Summary

- Features
 - Enable compact representation of grammatical constraints
 - Capture basic linguistic patterns
- Unification
 - Creates and maintains consistency over features
- Integration with parsing allows filtering of ill-formed analyses

HW #5

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Feature-based Parsing

- Goals:
 - Explore the role of features in implementing linguistic constraints.
 - Identify some of the challenges in building compact constraints to define a precise grammar.
 - Gain some further familiarity with NLTK.
 - Apply feature-based grammars to perform grammar checking.
- Individual work

Task

- Create grammar rules with features
 - Produce a single parse for grammatical sentences
 - Single parse per line
 - Reject ungrammatical sentences
 - Print blank line
- Homework includes sentences and “key”

Feature Grammar in NLTK

- NLTK supports feature-based grammars, including
 - ways of associating features with CFG rules
 - readers for feature grammars
 - *.fcfg* files
 - parsers
 - `Nltk.parse.FeatureEarleyChartParser`
- Nice discussion, examples in NLTK book CH. 9 (Ch. 8, ed1)
 - NOTE: HPSG-style comps list `<NP,PP,..>` NOT built into NLTK
 - Can be approximated with pseudo-list: e.g. `[FIRST=?a, REST=?b]`
 - For Extra-credit

Feature Structures

- `>>> fs1 = nltk.FeatStruct("[NUM='pl']")`
- `>>> print fs1`
- `[NUM='pl']`
- `>>> print fs1['NUM']`
- `pl`
- More complex structure
- `>>> fs2 = nltk.FeatStruct("[POS='N',
AGR=[NUM='pl',PER=3]]")`

Reentrant Feature Structures

- First instance
 - Parenthesized integer: (1)
- Subsequent instances:
 - 'Pointer': -> (1)
 - ```
>>> print(nltk.FeatStruct("[A='a', B=(1)[C='c'],
D->(1)]"))
```
  - [ A = 'a' ]
  - [ B = (1) [ C = 'c' ] ]
  - [ D -> (1) ]

# Augmenting Grammars

- Attach feature information to non-terminals, on
  - $N[AGR=[NUM='pl']] \rightarrow \text{'students'}$
  - $N[AGR=[NUM='sg']] \rightarrow \text{'student'}$
- So far, all values are literal or reentrant
  - Variables allow generalization: ?a
    - Allows underspecification, e.g.  $Det[GEN=?a]$
  - $NP[AGR=?a] \rightarrow Det[AGR=?a] N[AGR=?a]$

# Mechanics

- `>>> fs3 = nltk.FeatStruct(NUM='pl',PER=3)`
- `>>> fs4 = nltk.FeatStruct(NUM='pl')`
- `>>> print fs4.unify(fs3)`
- `[NUM = 'pl']`
- `[PER = 3 ]`