# Feature-Based Grammar

Ling571 Deep Processing Techniques for NLP February 4, 2015

# 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

#### • S -> NP VP

- S -> NP[NUM=?n] VP[NUM=?n]
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- Det[NUM=pl] -> 'these' | 'all'
- N-> 'dog' | 'girl' | 'car' | 'child'

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- N[NUM=pl] -> 'dogs' | 'girls' | 'cars' | 'children'

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- >>> 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

#### 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

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- Integrate unification in parse construction

# Parsing, Unification, & Earley

- Augment existing Earley parser for unification
  - Fairly straightforward
- Modify representations:
  - Augment CFG rules with constraints
    - Use constraints to create feature structure as DAG
  - Add DAG to state representation
    - E.g., S -> NP VP, [0,0],[],Dag

# Integrating Unification

- Main change: Completer
  - Advances in rules where next constituent matches a just-completed constituent
  - Now, unifies Dag from completed constituent with the part of the feature structure in rules advanced
    - If fails, no new entry in chart
- Second change:
  - Only add state if NOT subsumed by states in chart

#### Notes on Features

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#### Feature Grammar in NLTK

- NLTK supports feature-based grammars
  - Includes ways of associating features with CFG rules
  - Includes readers for feature grammars
    - *.fcfg* files
  - Includes parsers
    - NItk.parse.FeatureEarleyChartParser

## 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']] -> 'students'
  - N[AGR=[NUM='sg']] -> 'student'
- So far, all values are literal or reentrant
  - Variables allow generalization: ?a
    - Allows underspecification, e.g. Det[GEN=?a]
  - NP[AGR=?a] -> 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 ]

# 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:

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- E.g.:
  - ?The rocks slept.
  - ?Colorless green ideas sleep furiously.

- Many proposed:
  - Animacy: +/-
  - Natural gender: masculine, feminine, neuter
  - Human: +/-
  - Adult: +/-
  - Liquid: +/-
  - Etc.
  - The milk spilled.
  - ?The cat spilled.

#### Examples

- The climber hiked for six hours.
- The climber hiked on Saturday.
- The climber reached the summit on Saturday.
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- Contrast:
  - Achievement vs activity

# Semantic features & Parsing

- Can filter some classes of ambiguity
  - Old men and women slept.
  - (Old men) and (women) slept.
  - (Old (men and women)) slept.
  - Sleeping people and books lie flat.
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### 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 illformed analyses



Grammar entry for sentence



Grammar entry for NP

Cat	DT ]	
definite	yes	
number	SG	
form	"the"	
[ cat	DT	
definite	yes	
definite number	yes PL	

Lexical entries





Unifying NP with Determiner

