

Feature Structures and How to Represent Multiple Phenomena Simultaneously

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Outline

- Definitions and Examples
- Parsing with Feature Structures
- The Earley Algorithm
- Other Issues
- GLARF: a Feature Structure Project at NYU



Why Feature Structures?

- A Feature Structure is a good data structure for representing complex objects
 - Can include many linguistic features in one structure: Tense, Agreement, Semantics, Parsed Structure, Coreference, ...
- Represents objects in terms of features value pairs, where the values of features can be complex
- The mathematics of Feature Structures were worked out in great detail in the 1980s and 1990s
- Several linguistic theories are formalized in terms of Feature Structures and operations thereon



Defining Feature Structures

- A Feature Structure is either atomic or a set of feature value pairs
 - $FS \rightarrow NIL$
 - $FS \rightarrow Atom$
 - $FS \rightarrow \{FV_1, FV_2, \dots FV_N\}$
 - $FV \rightarrow Feature = FS$
 - **A values of a feature must be a FS**
- Each Feature and Value Represents a Piece of Information
- More information defines more specific objects



A Simple Example

- $FS_1 = [\text{Color} = \text{Green}]$
 - Describes a green thing
- $FS_2 = [\text{Height} = \text{Tall}]$
 - Describes a tall thing
- $FS_3 = [\text{Color} = \text{Green}, \text{Height} = \text{Tall}]$
 - Describes a tall green thing
- More feature value pairs describe a more specific thing



Typed Feature Structures

- Typed feature structures:
 - Every feature structure has a type
 - The type limits what are the possible features that can be included in it
 - Every feature has a type
 - The type limits its possible values
- Examples
 - A Feature Structure of type **Lego** allows features: color, height, width, depth and material.
 - The value of the feature **Color** allows atomic TFS as values from the set {**red**, **yellow**, **blue**, **green**, ...}



Subsumption

- The operator \sqsubseteq represents “subsumes”
- $FS_1 \sqsubseteq FS_2$, if FS_1 describes the same or larger set of possible entities than FS_2 does.
 - For example, if FS_1 represents something green and FS_2 represents a tall green thing, then $FS_1 \sqsubseteq FS_2$
 - **[Color = Green] \sqsubseteq [Color = Green, Height = Tall]**
- Notice that if $FS_1 \sqsubseteq FS_2$, then FS_2 includes all of the Feature Value pairs in FS_1 , but the reverse may not be true.
- For typed feature structures, one must add information about type subsumption and this is essentially based on the definitions of types (similar to type inheritance in OOP)
 - I will leave out some of the details about types, but can talk more about them if there are questions.

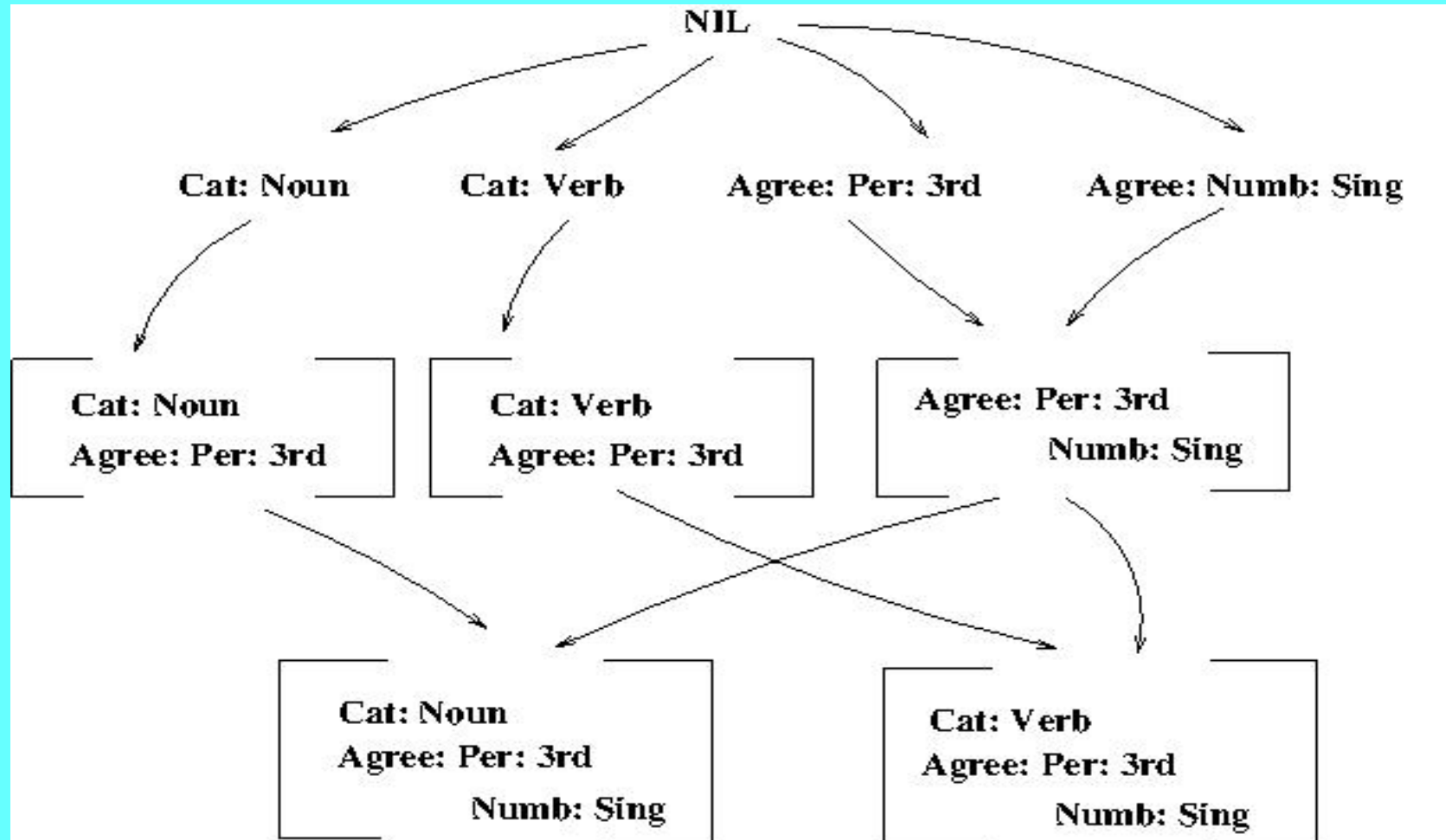


Properties of Subsumption

- NIL is the most general feature structure
 - Subsumes every other feature structure
 - The set of zero feature value pairs
 - Also subsumes atomic feature structure
 - Possible value for all features (for typed feature structures)
- Subsumption is transitive
- If $FS_1 \sqsubseteq FS_2$ and $FS_2 \sqsubseteq FS_3$, then $FS_1 \sqsubseteq FS_3$
- Subsumption partially orders the set of all FS
 - NIL is the root of a DAG which includes all FSs
 - Edges in paths from the root represent subsumption



Part of the Subsumption Graph for a FS-based Grammar of English



Unification

- Unifying (operator = \sqcup) two FSs combines the information in both feature structures to produce a FS that instantiates the intersection of entities that the two input FSs instantiate
- $FS_1 \sqcup FS_2 = FS_3$ iff FS_3 is the most general Feature structure (the one with the fewest Feature Value pairs) such that:
 - $FS_1 \sqsubseteq FS_3$ and $FS_2 \sqsubseteq FS_3$
- Properties:
 - Unification is Commutative
 - $FS_1 \sqcup FS_2 = FS_2 \sqcup FS_1$
 - Unification is Associative
 - $(FS_1 \sqcup FS_2) \sqcup FS_3 = FS_1 \sqcup (FS_2 \sqcup FS_3)$



How to Unify (not worrying about efficiency)

- $FS_X \sqcup NIL \rightarrow FS_X$
- $NIL \sqcup FS_X \rightarrow FS_X$
- $Atom_1 \sqcup Atom_2$ Fails if ***Atom₁ ≠ Atom₂***
- To Unify Complex FSs FS_1 and FS_2 , producing FS_3 , start with an empty FS_3 and add FVs as follows:
 - For each Feature Value Pair FV_1 in FS_1 , try to find a matching FV_2 in FS_2 such that Feature F_1 in FV_1 is the same as F_2 in FV_2
 - If no matching feature exists, then add FV_1 into FS_3
 - Otherwise, try to unify V_1 in FV_1 with V_2 in FV_2
 - If Fail, then unification fails
 - Otherwise, add F with a value of $V_1 \sqcup V_2$ to FS_3
 - Add all FVs in FS_2 that did not match any Feature in FS_1

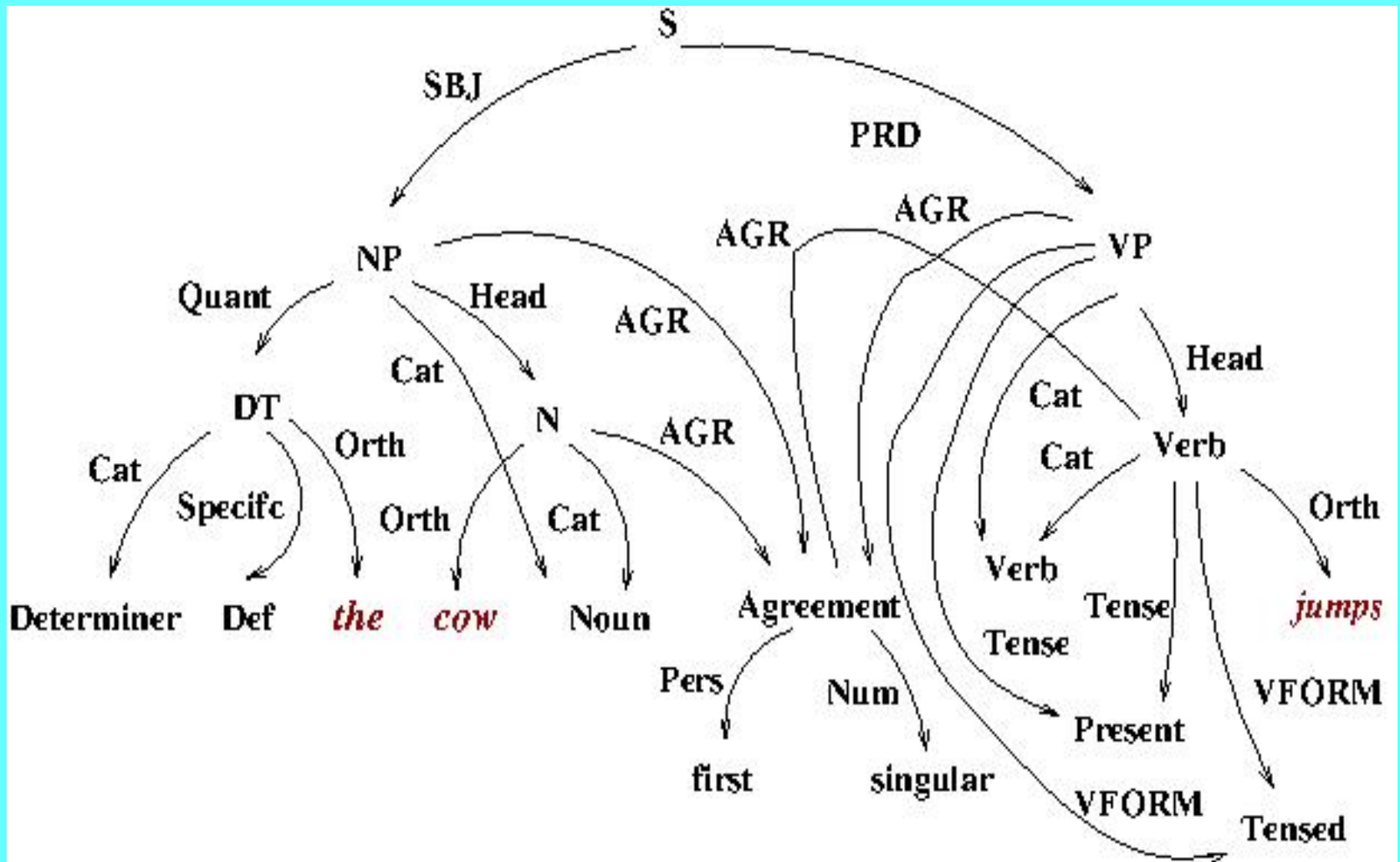


Feature Structures as Edge-Labeled DAGs

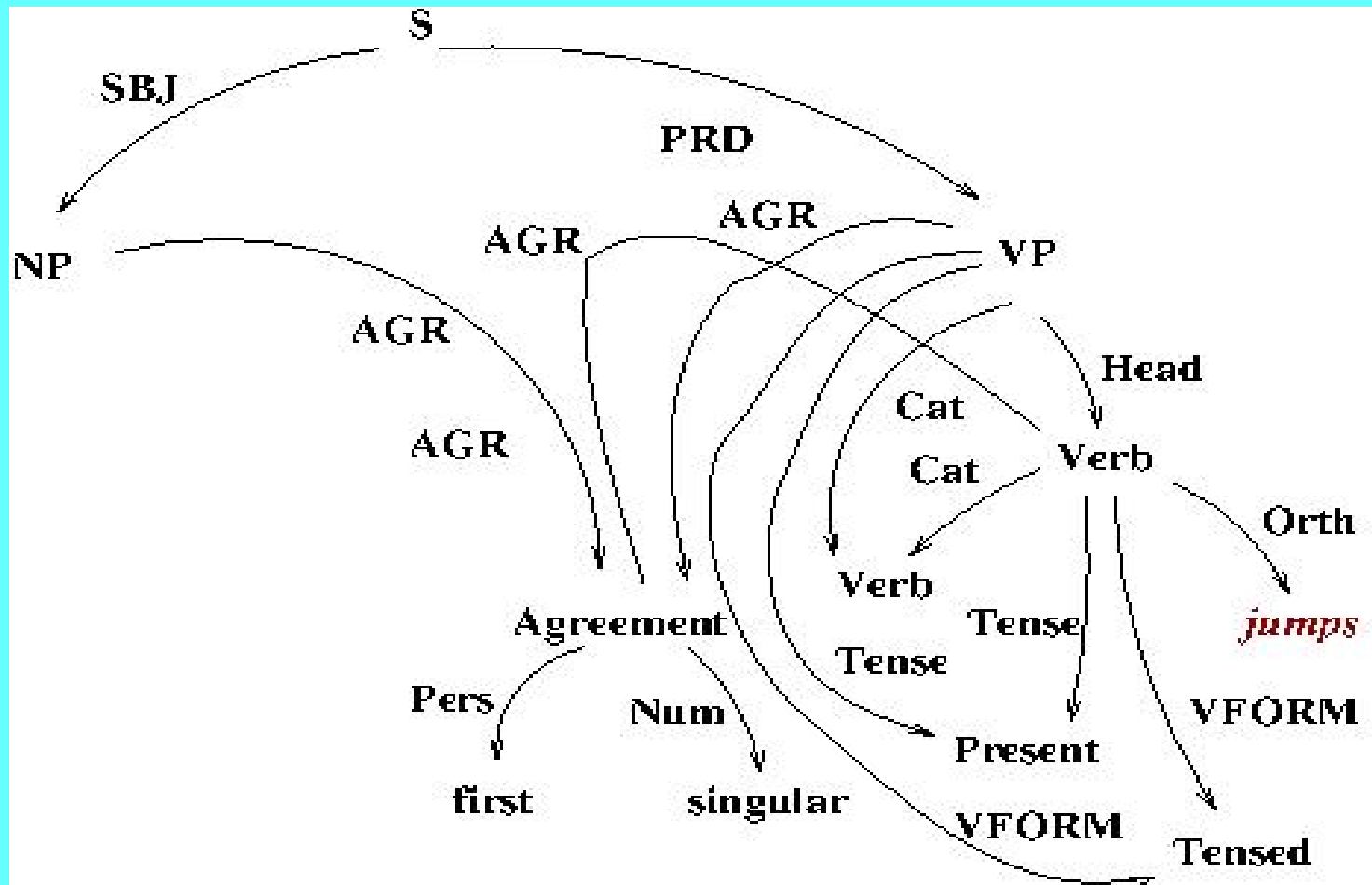
- Types = Internal Nodes = Non Terminals = Phrasal Categories and Parts of Speech
- Atomic FSs = leaves
- Features = Edge Labels
- Shared Structure is determined by grammar
 - It means that some features values are exactly the same
 - Common Instances
 - Shared between a phrase and its head
 - Agreement between a subject and a verb



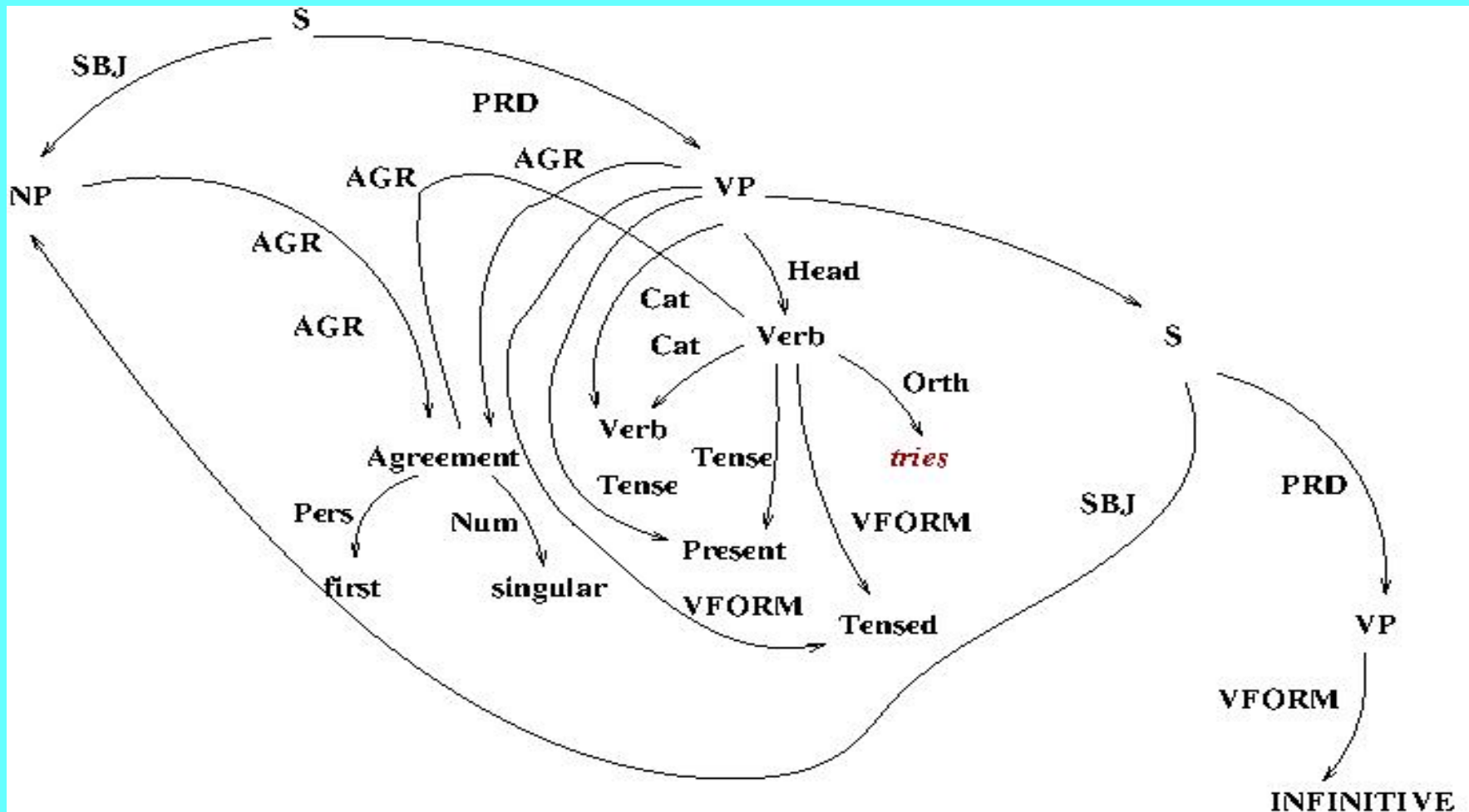
FS representing *The cow jumps*



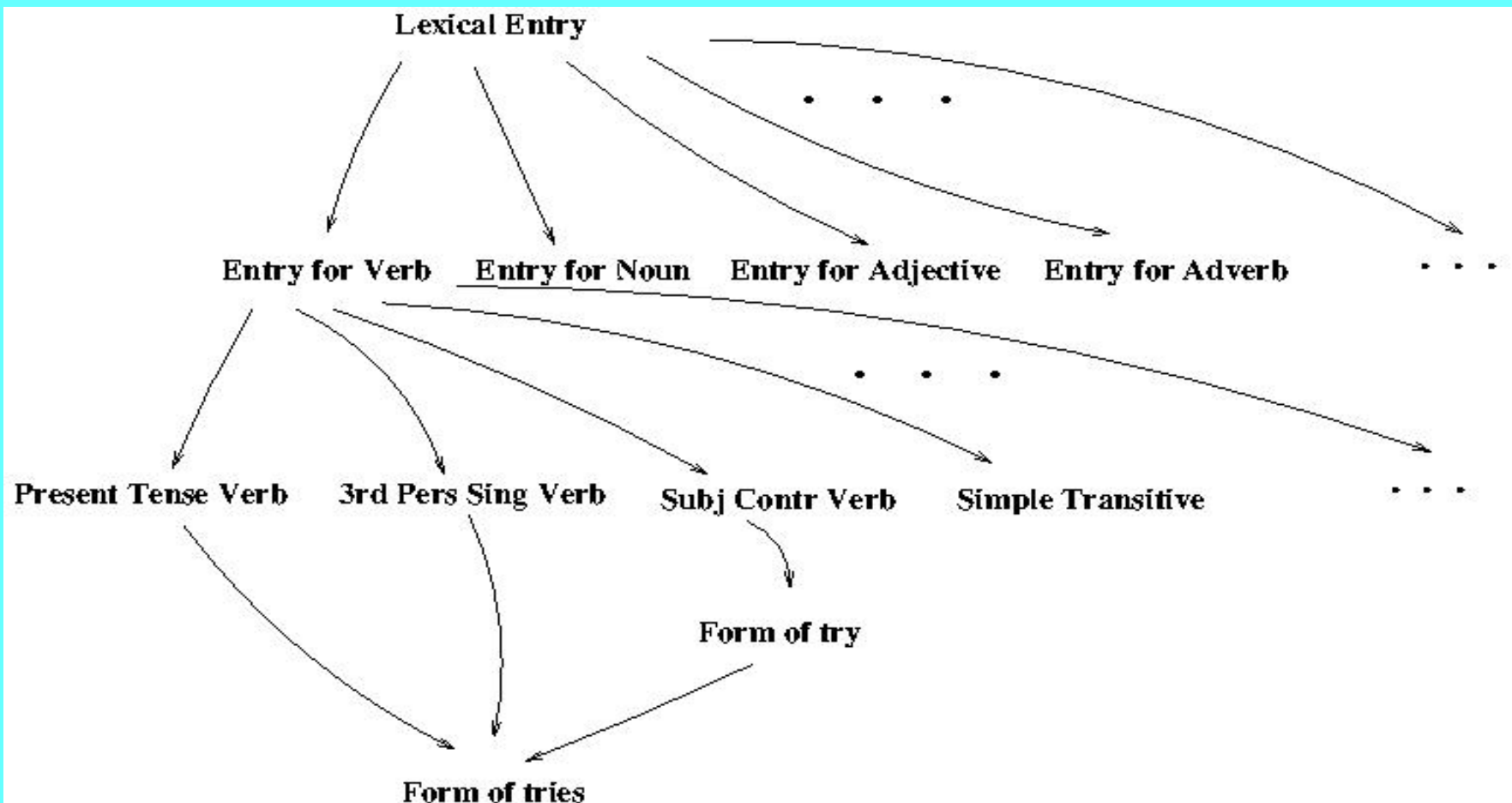
FS for lexical entry for *jumps*



FS Lexical Entry for the Verb *tries*



Lexicon Can Be Arranged Hierarchically, based on Subsumption



How Can We Use FSs for Parsing?

- For each word, we look up all its feature structure entries (instead of looking up its possible parts of speech)
 - These FSs or generalizations of these feature structures can correspond to either:
 - Initial Terminal Symbols, e.g., FS representing a noun
 - Initial NonTerminal Symbols, e.g., FS representing an S licensed by a verb
- Do we Need Context Free Grammars?
 - Using the second type of entries, it is possible to (in a way) fold the entire grammar into the lexicon
 - Alternatively, a context free grammar can be used to guide the combination of FSs, as in standard parsing
 - FSs constrain possible combinations



The Earley Algorithm

- Shortcoming of Top Down Parsing
 - Left Recursive rules like $\mathbf{NP} \rightarrow \mathbf{NP} \mathbf{PP}$
 - If \mathbf{NP} is recognized, productions starting with \mathbf{NP} are added to chart including this rule which starts with \mathbf{NP} (hence infinite recursion)
- The Earley Algorithm solves this problem:
 - it avoids adding duplicate productions to the chart
- Productions $\mathbf{XP} \rightarrow \mathbf{X}_1 \cdot \mathbf{X}_2 \mathbf{X}_3 [i,j]$ in the chart include:
 - A phrase structure rule ($\mathbf{XP} \rightarrow \mathbf{X}_1 \cdot \mathbf{X}_2 \mathbf{X}_3$)
 - A dot (between \mathbf{X}_1 and \mathbf{X}_2) such that complete constituents to the left of the dot have been matched
 - The span of text that this rule applies to between i and j
- The Earley algorithm would not add $\mathbf{NP} \rightarrow \mathbf{NP} \cdot \mathbf{PP} [0,1]$
 - If there was already an instance in the chart



FS version of the Earley Algorithm

- We assume the model in which phrase structure rules guide combination of FSs
 - A parsing step combines 1 complete and 1 incomplete states
 - A state is complete if the dot is all the way to the right
 - $XP \rightarrow X_1 X_2 X_3 \cdot$
 - An incomplete state has the dot somewhere else
 - $YP \rightarrow W_1 \cdot XP Z_3$
 - The result combines the two by matching the complete state with the symbol following the dot and then advancing the dot
 - $YP \rightarrow W_1 XP \cdot Z_3$
- For the FS version, matching is based on subsumption
 - Matching for purposes of a parsing step (above)
 - When checking if a production is already in the chart (previous slide)



Efficiency Issues for FS Parsing

- Efficient unification changes input FSs
 - Combining them destructively keeping parts of each
- For chart parsing, original FSs are needed
 - So FS parsing involves lots of copying (this can be inefficient)
- Solutions
 - Use general FSs in productions that subsume “real ones”
 - Generate final FS after final parse is found
 - Lazy copying (Godden 1990)
 - Use instruction like “copy FS₁” to delay copying
 - Then copy only when FS is actually needed



GLARF

- See CUNY talk



Readings

- J & M Chapters 13.4.2 and 15

