L95: Natural Language Syntax and Parsing 8) Unification-based Grammars and Parsing

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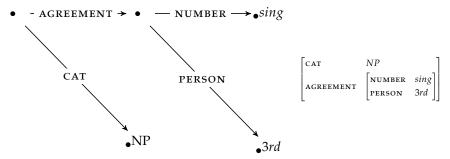
Reminder...

Last time we looked at lexicalisation and features to help us with:

- modelling structural dependency across the tree as a whole
 - e.g. correctly modelling *NP* expansion
- modelling the structural behaviour specific to a lexical item:
 - pp-attachment
 - subcategorisation
 - co-ordination

Alternative approach represents features in **DAGs**

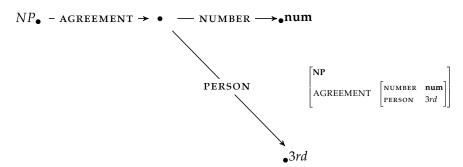
Re-conceptualise words, non-terminal nodes and parses as **Directed Acyclic Graphs** which may be represented as **Attribute Value Matrices**



We have atomic categories at each of the terminal nodes and another AVM/DAG at all other nodes

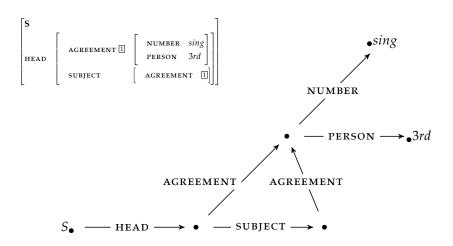
Some grammars allow the AVMs to be typed

Typing facilitates grammar building. Hierarchies of AVM types can be used to automatically populate features



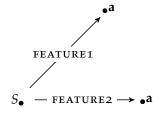
An shorthand notation uses angle bracket notation to indicate feature paths: e.g. < NP AGREEMENT PERSON> would represent the feature path leading to the atomic value 3rd.

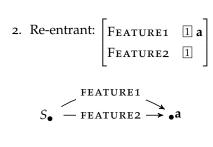
DAGs and AVMs may exhibit re-entrancy



DAGs and AVMs may exhibit re-entrancy

1. Non re-entrant: $\begin{bmatrix} F_{EATURE1} & a \\ F_{EATURE2} & a \end{bmatrix}$



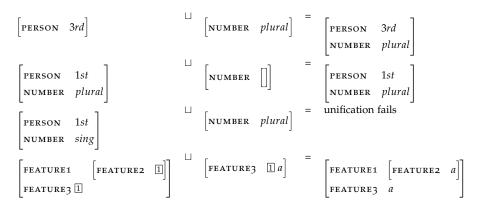


Parsing with DAGs involves Unification

- The unification of two DAGs is the most specific DAG which contains all the information in both of the original feature structures.
- Unification fails if the two DAGs contain conflicting information.

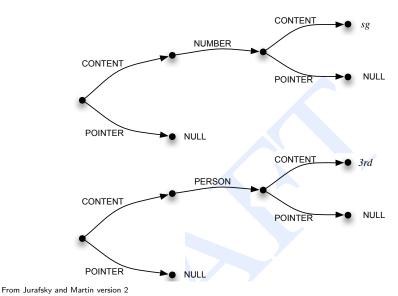
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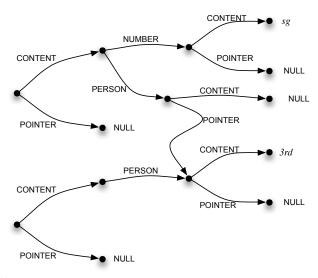


Unification examples in class

Unification algorithm requires extra graph structure

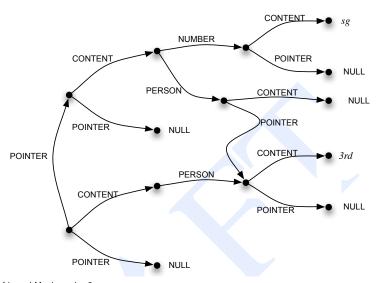


Unification algorithm requires extra graph structure



From Jurafsky and Martin version 2

Unification algorithm requires extra graph structure



From Jurafsky and Martin version 2

DAGs can be straighforwardly associated with the lexicon

```
{fish, rivers, pools, they}
AGREEMENT PERSON 3rd NUMBER plural
\left\langle \text{they,} \begin{vmatrix} \mathbf{N} \\ \mathbf{AGREEMENT} \end{vmatrix} \begin{vmatrix} \mathbf{PERSON} & 3rd \\ \mathbf{NUMBER} & sing \end{vmatrix} \right| \right\rangle
```

DAGs can be straighforwardly associated with the lexicon

```
{fish, rivers, pools, they}
AGREEMENT PERSON 3rd NUMBER plural
   \rightarrow {cans, fishes}
                \langle V | AGREEMENT PERSON \rangle = 3rd
                <V AGREEMENT NUMBER> = sing
\left\langle \text{they,} \begin{vmatrix} \mathbf{N} \\ \text{AGREEMENT} \end{vmatrix} \begin{vmatrix} \text{PERSON} & 3rd \\ \text{NUMBER} & sing \end{vmatrix} \right| \right\rangle
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DAGs can be straighforwardly associated with the lexicon

$$\begin{bmatrix} \mathbf{N} \\ \mathbf{AGREEMENT} & \begin{bmatrix} \mathbf{PERSON} & 3rd \\ \mathbf{NUMBER} & plural \end{bmatrix} \end{bmatrix} \rightarrow \{ \text{fish, rivers, pools, they} \}$$

$$\mathbf{V} \rightarrow \{ \{ \text{cans, fishes} \} \}$$

$$\mathbf{V} \rightarrow \{ \{ \text{Cans, fishes} \} \}$$

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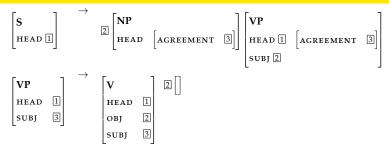
We can **modify** CFG algorithms to **parse** with DAGs

- We can use any CFG parsing algorithm if:
- associate feature constraints with CFG rules
- unify DAGs in the states
- $S \rightarrow NP VP$

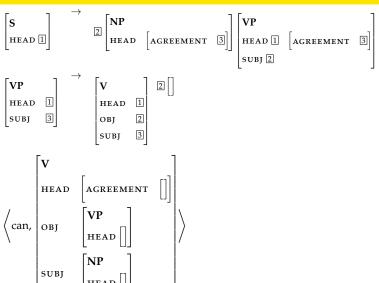
```
< NP HEAD AGREEMENT > = < VP HEAD AGREEMENT > < S HEAD > = < VP HEAD >
```

 We would have items like [X, [0, m], DAG] on the agenda or at each cell

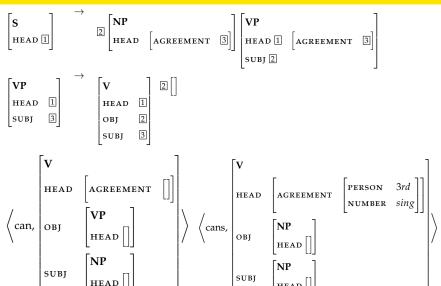
Subcategorization is captured by the feature constraints



Subcategorization is captured by the feature constraints



Subcategorization is captured by the feature constraints



Alternatively use unification as the parsing operation

Alternatively use **unification as the parsing operation** instead of just for search-space reduction through feature constraining:

•
$$X_0 \rightarrow X_1 X_2$$

 $< X_1$ HEAD AGREEMENT >=< X_2 HEAD AGREEMENT >
 $< X_0$ HEAD >=< X_1 HEAD >

- $X_0 \rightarrow X_1 X_2$ $< X_0$ HEAD $>< X_1$ HEAD > $< X_2$ CAT >= PP
- $X_0 \rightarrow X_1$ and X_2 $< X_0$ CAT $>< X_1$ CAT > $< X_1$ CAT $>< X_2$ CAT >

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$$X_0 \rightarrow X_1 X_2$$

 $< X_0 \text{ HEAD} >< X_1 \text{ HEAD} >$
 $< X_2 \text{ CAT} >= PP$

• $X_0 \rightarrow X_1$ and X_2 $< X_0$ CAT $>< X_1$ CAT > $< X_1$ CAT $>< X_2$ CAT >

Alternatively use unification as the parsing operation

Alternatively use **unification as the parsing operation** instead of just for search-space reduction through feature constraining:

$$ullet$$
 $X_0 o X_1 X_2$ $< X_1$ HEAD AGREEMENT $> = < X_2$ HEAD AGREEMENT $>$ $< X_0$ HEAD $> = < X_1$ HEAD $>$

•
$$X_0 \rightarrow X_1 X_2$$

 $< X_0 \text{ HEAD} >< X_1 \text{ HEAD} >$
 $< X_2 \text{ CAT} >= PP$

•
$$X_0 o X_1$$
 and X_2
 $< X_0$ CAT $>< X_1$ CAT $>$
 $< X_1$ CAT $>< X_2$ CAT $>$

Unification based parsing in the wild...

- Focus on adequacy for a wide range of languages as well as tractable for parsing
- Examples include Lexical Functional Grammar, LFG (Bresnan and Kaplan) and Head-driven Phrase Structure Grammar, HPSG (Pollard and Sag)
- Grammars tend to incorporate aspects of morphology, syntax and compositional semantics:

If you are interested see: http://www.delph-in.net