# Dependency Parsing & Feature-based Parsing

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Deep Processing Techniques for NLP
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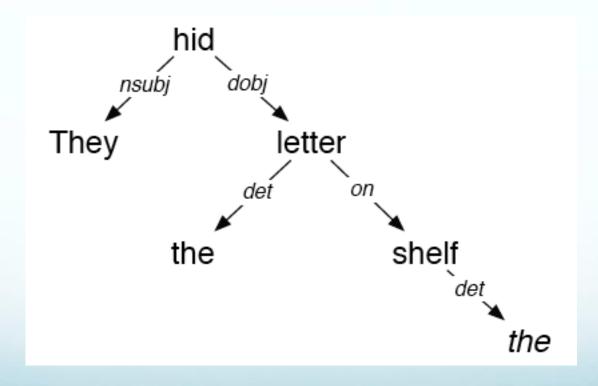
# Roadmap

- Dependency parsing
  - Transition-based parsing
    - Configurations and Oracles

- Feature-based parsing
  - Motivation
  - Features
  - Unification

# Dependency Parse Example

They hid the letter on the shelf



# Transition-based Parsing

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- Alternative methods for learning/decoding:
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  - Highest scoring: 2008, (and current)

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  - Most common model
    - Greedy classification-based approach
- Very efficient method: O(n)
- Some of the most successful systems:
  - Highest scoring: 2008, (and current)
- Best-known implementations:
  - Nivre's MALTParser (2006, and onward)

# **Transition Systems**

- A transition system for dependency parsing is:
  - A set of configurations
  - A set of transitions between configurations
  - An initialization function (for  $C_0$ )
  - A set of terminal configurations

# Configurations

• A configuration for a sentence x is the triple  $(\Sigma,B,A)$ :

- Σ is a stack with elements corresponding to the nodes (words + ROOT) in x
- B (aka the buffer) is a list of nodes in x
- A is the set of dependency arcs in the analysis so far,
  - $(w_i, L, w_i)$ , where  $w_x$  is a node in x, and L is a dep. label

#### **Transitions**

- Transitions convert one configuration to another s.t.
  - $C_i = t(C_{i-1})$ , where t is the transition

A dependency graph for a sentence is the set of arcs resulting from a sequence of transitions

The parse of the sentence is that resulting from the initial state through the sequence of transitions to a legal terminal state.

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- To parse a sentence, we need the sequence of transitions that derives it.
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# Dependencies -> Transitions

- To parse a sentence, we need the sequence of transitions that derives it.
- How can we determine sequence of transitions to parse a sentence?
  - Oracle: Given a dependency parse, identifies transition sequence.
    - Nivre's Arc-standard parser: provably sound & complete
  - Training:
    - Use oracle method on dependency treebank
      - Identifies gold transitions for configurations
    - Train classifier to predict best transition in new config.

- Words: w<sub>1</sub>,....,w<sub>n</sub>; Index 0: "dummy root"
- Initialization:
  - Stack = [0]; Buffer = [1,..,n]; Arcs = Ø

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  - Stack = [0]; Buffer = []; Arcs = A
- Transitions:
  - Shift: Push first element of buffer on top of stack
    - [i][j,k,,n][] → [i|j][k,...,n][]

- Transitions: Left-Arc label
  - Make left-arc between top two elements of stack
    - Remove dependent from stack, add arc to A
  - [i|j] [k,...,n] A  $\rightarrow$  [j] [k,...n] A U (j, label, i)
    - i not root

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  - Make left-arc between top two elements of stack
    - Remove dependent from stack, add arc to A
  - $[i|j] [k,...,n] A \rightarrow [j] [k,...n] A U (j, label, i)$ 
    - i not root
- Transitions: Right-Arc label
  - Make right-arc between top two elements of stack
    - Remove dependent from stack, add arc to A
  - $[i|j] [k,...,n] A \rightarrow [i] [k,...n] A U (i, label, j)$

- · Algorithm:
- Initialize configuration
- While (Buffer not empty) and (stack not only root):
  - If (top of stack is head of 2<sup>nd</sup> in stack) and (all children of 2<sup>nd</sup> attached), then Left-Arc
  - If (2<sup>nd</sup> in stack is head of top of stack) and (all children of top are attached), then Right-Arc
  - Else shift

[]

[They told him a story]

[] [They told him a story]
Shift [They] [told him a story]

Shift

Shift

П

[They]

[They, told]

[They told him a story]

[told him a story]

[him a story]

Shift

Shift

Left-Arc (subj)

Π

[They]

[They, told]

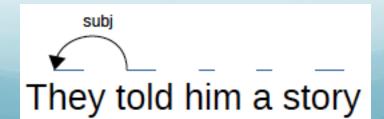
[told]

[They told him a story]

[told him a story]

[him a story]

[him a story)



Shift

Shift

Left-Arc (subj)

Shift

[]

[They]

[They, told]

[told]

[told, him]

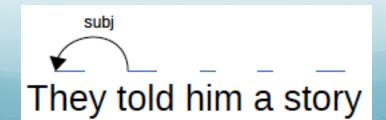
[They told him a story]

[told him a story]

[him a story]

[him a story)

[a story]



Shift

Shift

Left-Arc (subj)

Shift

Right-Arc (iobj)

[They]

[They, told]

[told]

[told, him]

[told]

[They told him a story]

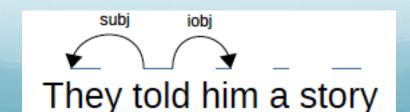
[told him a story]

[him a story]

[him a story)

[a story]

[a story]



Shift

Shift

Left-Arc (subj)

Shift

Right-Arc (iobj)

Shift

[They]

[They, told]

[told]

[told, him]

[told]

[told, a]

[They told him a story]

[told him a story]

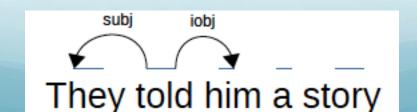
[him a story]

[him a story)

[a story]

[a story]

[story]



Shift

Shift

Left-Arc (subj)

Shift

Right-Arc (iobj)

Shift

Shift

[They]

[They, told]

[told]

[told, him]

[told]

[told, a]

[told, a, story]

[They told him a story]

[told him a story]

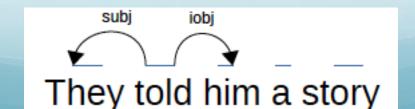
[him a story]

[him a story)

[a story]

[a story]

[story]



Shift

Shift

Left-Arc (subj)

Shift

Right-Arc (iobj)

Shift

Shift

Left-Arc (Det)

IJ

[They]

[They, told]

[told]

[told, him]

[told]

[told, a]

[told, a, story]

[told, story]

[They told him a story]

[told him a story]

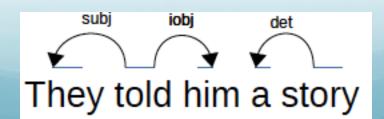
[him a story]

[him a story)

[a story]

[a story]

[story]



Shift

Shift

Left-Arc (subj)

Shift

Right-Arc (iobj)

Shift

Shift

Left-Arc (Det)

Right-Arc (dobj)

[They told him a story]

[They] [told him a story]

[They, told] [him a story]

[told] [him a story)

[told, him] [a story]

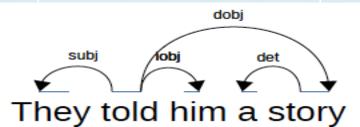
[told] [a story]

[told, a] [story]

[told, a, story]

[told, story] [

[told]



Is this a parser?

- Is this a parser?
- · No!
  - Uses known dependency tree information

- Creates transition-based representation of tree/parse
  - I.e. the sequence of transitions that derive it

# Transition-Based Parsing

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  - Can be viewed as finding highest scoring sequence
    - E.g. SH, LA, SH, RA, SH, SH, LA, RA
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## Transition-Based Parsing

- Find best sequence of transitions given input, model
  - · Can be viewed as finding highest scoring sequence
    - E.g. SH, LA, SH, RA, SH, SH, LA, RA
  - Where sequence score is sum of transition scores
  - Aka deterministic transition-based parsing
- Greedy approach: O(n) in length of input
  - Assuming O(1) access to highest scoring transition

# Greedy Transition-Based Parsing

- Parser  $(x = (w_1, ..., w_n))$ 
  - Initialize configuration c to start configuration on x
  - While not in terminal configuration:
    - Find best transition  $t^*$  on configuration c
    - Update c based on  $t^*$
  - Return set of output arcs (A)

How do we pick the best transition?

### **Transition Classification**

- How do we pick the best transition?
- Train classifier to predict transitions:
  - {Left-Arc,Right-Arc} x Number of dependency labels

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  - Original work: SVMs; Currently: DNNs + LSTM
- State-of-art dependency parsing:
  - UAS: 92.5%; LAS: 90.3%

# Building Transition-based Parser

- Given a dependency treebank
  - Use oracle algorithm to create transition sequences
  - Train classifier on transition x configuration features
  - Build greedy algorithm based on classifications
  - Evaluate on new sentences

# Dependency Parsing

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  - Lexicalized, localized
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- Dependency grammars:
  - Compactly represent pred-arg structure
  - Lexicalized, localized
  - Natural handling of flexible word order
- Dependency parsing:
  - Conversion to phrase structure trees
  - Graph-based parsing (MST), efficient non-proj O(n²)
  - Transition-based parser
    - MALTparser: very efficient O(n)
      - Optimizes local decisions based on many rich features

#### Features

#### Roadmap

- Features: Motivation
  - Constraint & compactness
- Features
  - Definitions & representations
- Unification
- Application of features in the grammar
  - Agreement, subcategorization
- Parsing with features & unification
  - Augmenting the parser, unification parsing
- Extensions: Types, inheritance, etc
- Conclusion

## Constraints & Compactness

- Constraints in grammar
  - $S \rightarrow NP VP$ 
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    - They run.
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  - But...
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  - Violate agreement (number), subcategorization

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      - $VP \rightarrow Vintrans$ ,
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  - Explosive!, loses key generalizations

- Need compact, general constraints
  - $S \rightarrow NPVP$

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    - Number, person, gender, etc

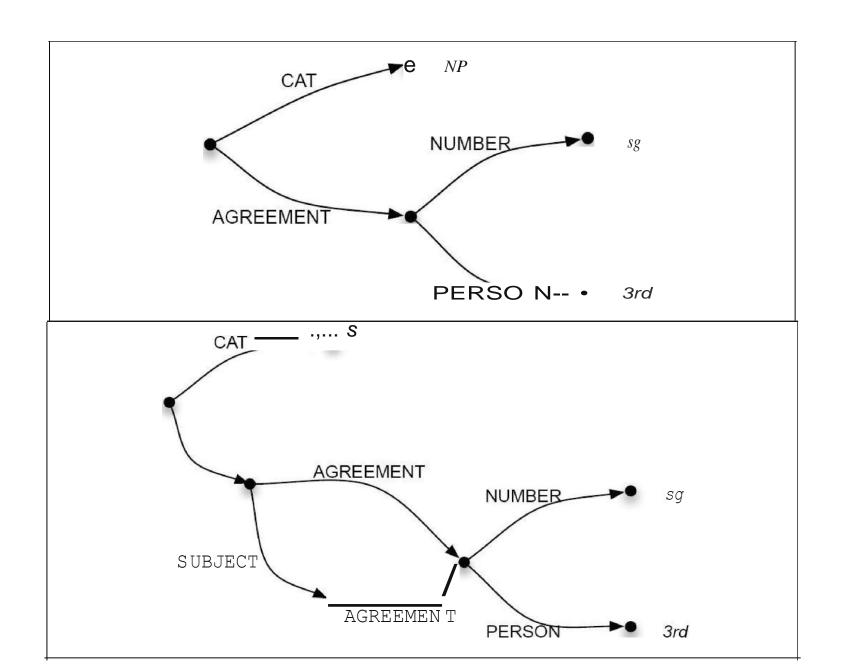
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  - E.g. Agreement
    - Number, person, gender, etc
- Augment CF rules with feature constraints
  - Develop mechanism to enforce consistency
  - Elegant, compact, rich representation

#### Feature Representations

- Fundamentally, Attribute-Value pairs
  - Values may be symbols or featurestructures
    - Feature path: list of features in structure to value
    - "Reentrant feature structures": share some struct
  - Represented as
    - Attribute-value matrix (AVM), or
    - Directed acyclic graph (DAG)

#### **AVM**

CAT NP **NUMBER** PL **NUMBER** PL**AGREEMENT PERSON** 3 **PERSON NUMBER** PL CAT 3 **PERSON NUMBER** PLAGREEM'T 1 **HEAD PERSON** CAT NP **AGREEMENT NUMBER** PL **SUBJECT PERSON** 3



Two key roles:

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- Two structures can unify if
  - Feature structures are identical
    - Result in same structure
  - Feature structures match where both have values, differ in missing or underspecified
    - Resulting structure incorporates constraints of both

## Subsumption

- Relation between feature structures
  - Less specific f.s. subsumes more specific f.s.
  - F.s. F subsumes f.s. G iff
    - For every feature x in F, F(x) subsumes G(x)
    - For all paths p and q in F s.t. F(p)=F(q), G(p)=G(q)

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- Examples:
  - A: [Number SG], B: [Person 3]
  - C:[Number SG]
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- Examples:
  - A: [Number SG], B: [Person 3]
  - C:[Number SG]
    - [Person 3]
  - A subsumes C; B subsumes C; B,A don't subsume
    - Partial order on f.s.

## **Unification Examples**

- Identical
  - [Number SG] U [Number SG]

- Identical
  - [Number SG] U [Number SG]=[Number SG]
- Underspecified
  - ' [Number SG] U [Number []]

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  - [Number SG] U [Number SG]=[Number SG]
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  - [Number SG] U [Number []] = [Number SG]
- Different specification
  - [Number SG] U [Person 3]

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- Different specification
  - [Number SG] U [Person 3] = [Number SG]
  - [Person 3]
  - [Number SG] U [Number PL]

- Identical
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- Underspecified
  - ' [Number SG] U [Number []] = [Number SG]
- Different specification
  - [Number SG] U [Person 3] = [Number SG]
  - [Person 3]
- Mismatched
  - [Number SG] U [Number PL] → Fails!

# More Unification Examples

```
AGREEMENT [1] U
SUBJECT AGREEMENT [1]
```

# More Unification Examples

```
AGREEMENT [1]

SUBJECT AGREEMENT [1]
```

```
SUBJECT AGREEMENT PERSON 3 NUMBER SG =

AGREEMENT [1]

SUBJECT AGREEMENT [1] PERSON 3 NUMBER SG
```

# 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

# Agreement with Heads and Features

```
VP \rightarrow Verb NP
<VP HEAD> = <Verb HEAD>
NP \rightarrow Det Nominal
\langle NP | HEAD \rangle = \langle Nominal | HEAD \rangle
<Det HEAD AGREEMENT> = <Nominal HEAD AGREEMENT>
Nominal → Noun
<Nominal HFAD> = <Noun HFAD>
Noun → flights
<Noun HEAD AGREEMENT NUMBER> = PL
Verb → serves
<Verb HEAD AGREEMENT NUMBER> = SG
<Verb HEAD AGREEMENT PERSON> = 3
```

# 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 Parsing**

- Employ constraints to restrict addition to chart
- Actually pretty straightforward

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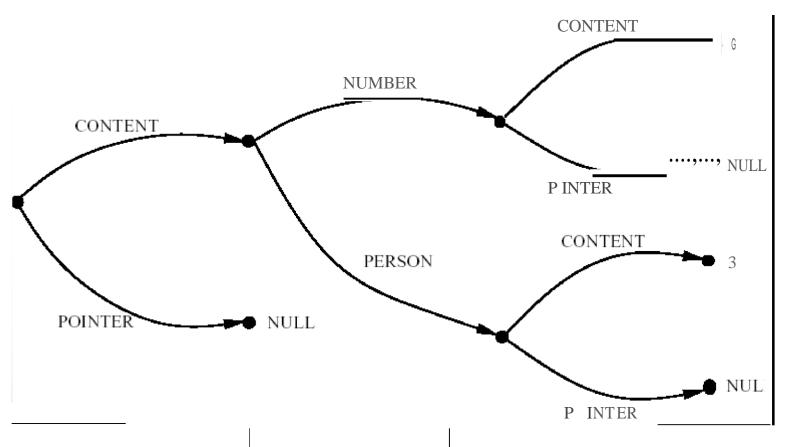
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#### Unification and Parsing

- 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

# Implementing Unification

- Data Structure:
  - Extension of the DAG representation
  - Each f.s. has a content field and a pointer field
    - If pointer field is null, content field has the f.s.
    - If pointer field is non-null, it points to actual f.s.



NUMBER PERSON SG 3

# Implementing Unification: II

- Algorithm:
  - Operates on pairs of feature structures
    - Order independent, destructive
  - If fs1 is null, point to fs2
  - If fs2 is null, point to fs1
  - If both are identical, point fs1 to fs2, return fs2
    - Subsequent updates will update both
  - If non-identical atomic values, fail!

# Implementing Unification: III

- If non-identical, complex structures
  - Recursively traverse all features of fs2
  - If feature in fs2 is missing in fs1
    - Add to fs1 with value null
  - If all unify, point fs2 to fs1 and return fs1

### Example

```
NUMBER SG

AGREEMENT [1]
     AGREEMENT [1]
     SUBJECT
SUBJECT AGREEMENT PERSON
  [ AGREEMENT [1]] U [AGREEMENT [PERSON 3]]
  [NUMBER SG] U [PERSON 3]
  [NUMBER SG] U [PERSON 3]
  [PERSON NULL]
```

#### Conclusion

- Features allow encoding of constraints
  - Enables compact representation of rules
  - Supports natural generalizations
- Unification ensures compatibility of features
  - Integrates easily with existing parsing mech.
- Many unification-based grammatical theories

# **Unification Parsing**

- Abstracts over categories
  - S-> NP VP =>
    - X0 -> X1 X2; <X0 cat> = S; <X1 cat>=NP;
    - <X2 cat>=VP
  - Conjunction:
    - $\times$  X0->X1 and X2; <X1 cat> =<X2 cat>;
    - < <X0 cat>=<X1 cat>
- Issue: Completer depends on categories
- Solution: Completer looks for DAGs which unify with the just-completed state's DAG

#### Extensions

- Types and inheritance
  - Issue: generalization across feature structures
    - E.g. many variants of agreement
      - More or less specific: 3<sup>rd</sup> vs sg vs 3rdsg
  - Approach: Type hierarchy
    - Simple atomic types match literally
    - Multiple inheritance hierarchy
      - Unification of subtypes is most general type that is more specific than two input types
    - Complex types encode legal features, etc

