#### Parsers

#### Introduction to Computational Linguistics: Parsing Algorithms

Sharon Goldwater (based on slides by Mark Steedman and Philipp Koehn)

20 July 2015

## informatics

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#### **Ambiguity refresher**

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Parsers need to handle (rampant!) syntactic ambiguity in natural language.

- **global ambiguity**: multiple full parses are possible, e.g., PP attachment: I saw the man with the telescope
- local ambiguity: ambiguous partial structures, need not be consistent with full parse.
  - classic garden path sentences: the old man the boats
  - but also lots of "normal" sentences: the dog bit the child
- Ambiguity can be structural (different possible phrasal constituents) or lexical (word with multiple POS tags), often both.

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# Parser properties

All parsers have two fundamental properties:

- Directionality: the sequence in which the structures are constructed.
  - top-down: start with root category (S), choose expansions, build down to words.
  - bottom-up: build subtrees over words, build up to S.
  - Mixed strategies also possible (e.g., left corner parsers)
- Search strategy: the order in which the search space of possible analyses is explored.

A **parser** is an algorithm that computes a structure for an input string given a grammar.

Understanding different parsing algorithms is important for:

- Computer scientists: parsers used to compile programs, check html, etc.
- NLP researchers: efficient parsers needed for large-scale language tasks (e.g., used to create Google's "infoboxes").
- Psycholinguists: what algorithm might be used by the human sentence processing mechanism?

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#### Parsing CFG refresher

Parsing algorithms exist for many types of grammars, but we'll consider just context-free grammars for now. CFG refresher (or see J&M 12.2-12.5):

- Two types of grammar symbols:
  - **terminals** (t): words.
  - Non-terminals (NT): phrasal categories like S, NP, VP, PP. Sometimes we distinguish pre-terminals (POS tags), a type of NT.
- Rules must have the form NT  $\rightarrow \beta$ , where  $\beta$  is any string of NT's and t's.
- A CFG in Chomsky Normal Form only has rules of the form  $NT_i \rightarrow NT_j NT_k$  or  $NT_i \rightarrow t_j$

Example: search space for top-down parser

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Start with S node.

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- Choose one of many possible expansions.
- Each of which has children with many possible expansions...
- etc



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#### Search strategies

- **depth-first search**: explore one branch of the search space at a time, as far as possible. If this branch is a dead-end, parser needs to **backtrack**.
- breadth-first search: expand all possible branches in parallel (or simulated parallel). Requires storing many incomplete parses in memory at once.
- **best-first search**: score each partial parse and pursue the highest-scoring options first. (Will get back to this when discussing statistical parsing.)

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### **Recursive Descent Parsing**

- A recursive descent parser treats a grammar as a specification of how to break down a top-level goal (find S) into subgoals (find NP VP).
- It is a top-down, depth-first parser:
  - blindly expand nonterminals until reaching a terminal (word).
  - If terminal matches next input word, continue; else, backtrack.

#### **Backtrack points**

- Need to keep track of backtrack points, to return to if we backtrack.
- Each backtrack point stores:
- A partial parse tree (what was completed when we made a choice)
- The rules we haven't tried yet
- The input words we haven't matched yet
- To ensure depth-first search, backtrack points are stored in a stack: last in, first out.

#### **RD** Parsing: initialization

We start with

• The rules of our context-free grammar, e.g.,

$\mathtt{S}  ightarrow \mathtt{NP}$ $\mathtt{VP}$	$\mathtt{VP} \to \mathtt{V}$	$\texttt{NN} \to \texttt{bit}$	$\mathtt{V}  ightarrow \mathtt{bit}$
$NP \rightarrow DT NN$	$DT \rightarrow the$	$NN \rightarrow dog$	$V \rightarrow dog$

- Current partial parse (also current subgoal): the S node.
- An ordered list of subgoals, initially containing just S.
- An empty stack of backtrack points.
- The input sequence (e.g., the dog bit)

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#### Parsing **RD** Parsing: iterative steps

- If first subgoal in list is a non-terminal A:
  - Pick a rule from the grammar to expand it (e.g.,  $A \rightarrow B C$ )
  - Replace A in subgoal list with B C
- If first subgoal in list is a *terminal* w:
  - If input is empty, backtrack.
  - If next input word is different from w, backtrack.
  - If next input word is w, match! i.e., consume input word w and subgoal w and move to next subgoal.

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#### **RD** Parsing: iterative steps

- If first subgoal in list is a non-terminal A:
  - Pick a rule from the grammar to expand it (e.g.,  $A \rightarrow B C$ )
  - Replace A in subgoal list with B C
- If first subgoal in list is a *terminal* w:
  - If input is empty, backtrack.\*
  - If next input word is different from w, backtrack.
  - If next input word is w, match! i.e., consume input word w and subgoal w and move to next subgoal.\*\*

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**Parsers vs Recognizers** 

• The above sketch is actually a recognizer: it tells us whether the sentence

Would need to add more details to keep track of parse structure as it is built.

- \* If stack is empty, we lose! No parse is possible.
- \*\* If no more subgoals, we win! We found a parse.

has a valid parse, but not what the parse is.

## **Recursive descent example**

	Stop	On	Subroals	Input
	Step	Op.	Jubgoals	input
	0		S	the dog bit
	1	Е	NP VP	the dog bit
• Grammar and sentence from	2	Е	DT NN VP	the dog bit
slide 9.	3	Е	the NN VP	the dog bit
	4	М	NN VP	dog bit
<ul> <li>Operations:</li> </ul>	5	Е	bit VP	dog bit
- Expand (E)	6	B4	NN VP	dog bit
- Match (M)	7	Е	dog VP	dog bit
= Backtrack to step $n$ (B $n$ )	8	М	VP	bit
Dackflack to step n (Dn)	9	Е	V	bit
	10	Е	bit	bit
	11	М		

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#### Shift-Reduce Parsing

- Search strategy and directionality are orthogonal properties.
- Shift-reduce parsing is depth-first (like RD) but bottom-up (unlike RD).
- Basic shift-reduce recognizer repeatedly:
  - Shifts input symbols onto a stack.
  - Whenever possible, reduces one or more items from top of stack that match RHS of rule, replacing with LHS of rule.
- Like RD parser, needs to maintain backtrack points.

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#### Shift-reduce example

	Step	Op.	Stack	Input
	0			the dog bit
	1	S	the	dog bit
	2	R	DT	dog bit
• Same example grammar and	3	S	DT dog	bit
sentence.	4	R	DT V	bit
	5	S	DT V bit	
<ul> <li>Operations:</li> </ul>	6	R	DT V V	
- Shift (S)	7	B5	DT V bit	
- Reduce (R)	8	R	DT V NN	
- Backtrack to step $n$ (B $n$ )	9	B3	DT dog	bit
	10	R	DT NN	bit
	11	R	NP	bit
	12	R	NP bit	bit

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#### Depth-first parsing in practice

- Depth-first parsers are very efficient for unambiguous structures.
  - Widely used to parse/compile programming languages
  - Language/grammar is specially constructed to be unambiguous (sometimes with finite lookahead).

Parsing Breadth-first search using dynamic programming

• With a CFG, a parser should be able to avoid re-analyzing sub-strings because

The dog saw a man in the park NP

NP

the analysis of any sub-string is *independent* of the rest of the parse.

NP

## Depth-first parsing in practice

- Depth-first parsers are very efficient for unambiguous structures.
  - Widely used to parse/compile programming languages
  - Language/grammar is specially constructed to be unambiguous (sometimes with finite lookahead).
- But can be massively inefficient (exponential in sentence length) if faced with local ambiguity.
  - Blind backtracking may require re-building the same structure over and over.
  - So, much less common for natural language parsing (though some work on best-first probabilistic shift-reduce parsing: uses lookahead to help predict which expansions to make).



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### Parsing Breadth-first search using dynamic programming

• With a CFG, a parser should be able to avoid re-analyzing sub-strings because the analysis of any sub-string is *independent* of the rest of the parse.

The	dog	saw	а	man	in	the	park
N	IP			NP		1	٧P
						Ы	

- To exploit this fact, chart parsing algorithms use dynamic programming to store and reuse sub-parses.
- This permits exploring multiple potential parses at once: a breadth-first strategy

Parsing **Depicting a WFST/Chart** 

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#### Parsing as dynamic programming

- As in HMM algorithms, dynamic programming fills a table of solutions to subproblems (memoization), then composes these to find the full solution.
- For parsing, subproblems are analyses of substrings, memoized in chart (aka well-formed substring table, WFST).
- · Chart entries are indexed by start and end positions in the sentence, and correspond to:
- either a complete constituent (sub-tree) spanning those positions (if working bottom-up).
- or a hypothesis about what complete constituent might be found (if working top-down).

Sharon Goldwater Parsing 20 Depicting a WFST as a Matrix 3 v 0 1 Example, partway through parsing. Prep PP Here, only showing root nodes of 2 Det NP each constituent. 3 Ν Lower left of chart never used; often only upper right is shown. 4 5  $_0$  See  $_1$  with  $_2$  a  $_3$  telescope  $_4$  in  $_5$  hand  $_6$ 

- Chart can be depicted as either a matrix or a graph.
  - In either case, we assume indices between each word in the sentence:

 $_0$  See 1 with 2 a 3 telescope 4 in 5 hand 6

- If using a matrix, cell [i, j] holds information about the word span from position *i* to position *j*:
  - The root node of any constituent(s) spanning those words
  - Pointers to its sub-constituents
  - (Depending on parsing method,) predictions about what constituents might follow the substring.

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#### Depicting a WFST as a Graph

- Here, each sentence position index is a node or vertex.
- edges (arcs) represent spans, labelled with the same information that goes in a cell in the matrix representation.



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#### **Algorithms for Chart Parsing**

Many different chart parsing algorithms, including

- the CKY algorithm, which memoizes only complete constituents
- · various algorithms that also memoize predictions/partial constituents
- often using mixed bottom-up and top-down approaches, e.g., the Earley algorithm described in J&M, and left-corner parsing.

#### **CKY Algorithm**

CKY (Cocke, Kasami, Younger) is a **bottom-up**, **breadth-first** parsing algorithm.

- Original (simplest) version assumes grammar in Chomsky Normal Form.
- Add constituent A in cell (*i*, *j*) if:
  - there is a rule  $A \rightarrow B$ , and B is in cell (i, j), or
  - there is a rule  $A \rightarrow B$  C, and B is in cell (i,k) and C is in cell (k,j).

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#### **CKY** Algorithm

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  - there is a rule  $A \rightarrow B$  C, and B is in cell (i,k) and C is in cell (k, j).
- Fills chart in order: only looks for rules that use a constituent from *i* to *j* after finding all constituents ending at i. So, guaranteed to find all possible parses.

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#### Parsing **CKY** Pseudocode

• Assume input sentence with indices 0 to n, and chart c.

```
for len = 1 to n:
                    #number of words in constituent
   for i = 0 to n-len:
                         #start position
      j = i+len #end position
      #process unary rules
      if A->B and c[i,j] has B, add A to c[i,j]
      for k = i+1 to j-1 #mid position
         #process binary rules
         if A->B C and c[i,k] has B and c[k,j] has C, add A to c[i,j]
```

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Visualizing the Chart 2

3

4

fish

• This algorithm performs recognition in time  $O(n^3)$ .

0

1

2

3

the

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Parsing **CKY** example

$\begin{array}{l} eq:sphere$	Lexical rules Det $\rightarrow$ a   the (determiner) N $\rightarrow$ fish   frogs   soup (noun) Prep $\rightarrow$ in   for (preposition) TV $\rightarrow$ saw   ate (transitive verb) IV $\rightarrow$ fish   swim (intransitive verb) Relpro $\rightarrow$ that (relative pronoun)
$VP \rightarrow TV NP$ $VP \rightarrow IV PP$	$Relpro \to that \ \textbf{(relative pronoun)}$
$VP\toIV$	
$PP \to Prep \; NP$	
SRel  o Relpro VP	

Nom: nominal (the part of the NP after the determiner, if any). SRel: subject relative clause, as in the frogs that ate fish.



Unary branching rules: det  $\rightarrow$  the

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ate

froas



Unary branching rules: N  $\rightarrow$  frogs, Nom  $\rightarrow$  N, NP  $\rightarrow$  Nom

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Unary branching rules:  $\mathsf{tv}\to\mathsf{ate}$ 







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Visualizing the Chart (1,4)

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Binary rule:  $S \rightarrow NP VP$  (1,2) & (2,4)  $\rightsquigarrow$  (1,4)

(1,3) & (3,4) 1

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#### Visualizing the Chart (3,4)



Unary branching rules: N  $\rightarrow$  fish, Nom  $\rightarrow$  N, NP  $\rightarrow$  Nom, iv  $\rightarrow$  fish, vp  $\rightarrow$  iv



#### (1,2) & (2,3) $\not \rightarrow$

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(0,1) & (1,3)  $\not \rightarrow$ 



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## Visualizing the Chart (0,4)



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(0,1) & (1,4)  $\not\sim$ Binary rule: S  $\rightarrow$  NP VP (0,2) & (2,4)  $\rightsquigarrow$  (0,4)

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#### A note about CKY ordering

- Notice that to fill cell (i, j), we use a cell from row i and a cell from column j.
- So, we must fill in all cells down and left of (i, j) before filling (i, j).
- Here, we filled in all short entries, then longer ones, but other orders can work (e.g., J&M fill in all spans ending at j, then increment j.)

## From CKY Recognizer to CKY Parser

- As just specified, CKY only recognizes, but can't return the parse.
- e.g., we don't know from S cell how it was constructed.
- Easy to fix:

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- whenever a constituent is found for cell (i,j), store the indices of the sub-constituents that formed it.
- can mean storing multiple copies of A with different indices.
- Sometimes called a packed parse forest: represents a possibly exponential number of trees in a compact way.

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#### **CKY** in practice

- Avoids re-computing substructures, so much more efficient than depth-first parsers (in worst case).
- Still may compute a lot of unnecessary partial parses.
- Simple version requires converting the grammar to CNF (may cause blowup).

Various other chart parsing methods avoid these issues by combining top-down and bottom-up approaches (see J&M).

We also haven't said anything about how to choose between different parses when there's global ambiguity.

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#### **Recap:** Parsing

We have seen several different parsing algorithms:

- Recursive descent parsing: top-down, depth-first.
- Shift-reduce parsing: bottom-up, depth-first.
- CKY parsing: bottom-up, breadth-first.

Do any of these seem plausbile as a model of human sentence processing?

Introduction to Computational Linguistics: Parsing and Human Sentence Processing

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20 July 2015

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### Human Parsing Properties of human parsing mechanism

• Fast: real-time.

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- Recognizes global ambiguity: at least to some extent.
- Incremental: words (and meaning) are integrated into structure immediately.
- Can be led astray: by local ambiguity (garden path sentences).
- But mostly isn't: many local ambiguities don't cause problems.

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## Human Parsing

- Global ambiguity
- Some examples are clearly (even humorously) ambiguous:
  - I saw the man with a telescope.
- She sat on the chair covered in dust.
- Milk drinkers are turning to powder.
- But most ambiguity isn't even noticed!
  - I saw the man with a hat.
  - She stood on the stoop covered in tears.
  - Breast feeders are turning to a new enriched formula.

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#### Local ambiguity

- Same goes for local ambiguity:
  - The old man the boats
  - We painted the wall with cracks
  - Fat people eat accumulates

versus

- The dog bit the cat
- The green is used for playing soccer
- We stopped short of going

#### Properties of our parsing algorithms

#### Serial versus parallel parsing

- Fast? We argued that RD and SR are inefficient, but could perhaps be improved by good heuristics for choosing next rules. Anyway, hard to evaluate what counts as "fast".
- Recognize global ambiguity? CKY builds all parses, so definitely ves. RD/SR can return multiple parses if run past the first one. But notice a distinction...
- Depth-first parsers are inherently serial: one structure processed at a time. So, recognizing ambiguity implies backtracking/re-parsing, and one structure will always be recognized first.
- Breadth-first parsers are idealized as parallel: multiple structures processed simultaneously. If truly parallel, ambiguous structures identified simultaneously.

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Human Parsing

#### Human parsing: serial or parallel?

- On the face of it, full parallel parsing seems implausible:
  - Finds/notices all the global ambiguities.
  - Doesn't get stuck in garden paths.
- Serial parsing provides a possible explanation for garden paths (backtracking).
- But there are parallel options too:
  - limited parallelism: pursue a fixed (small) number of structures at once, may still require occasional backtracking.
  - ranked parallelism: (possibly in combination with above). Possible structures ranked by preference; garden path if low-ranked structure turns out to be correct.

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#### Is CKY parsing incremental?

- The chart-filling order we used (short spans first) clearly isn't.
- What about J&M's ordering (fill all cells ending at j, then j+1)?
- Consider processing The girl gave the dog a bone:

S	$\rightarrow$	NP VP	Det	$\rightarrow$	the   a
NP	$\rightarrow$	Det CN	CN	$\rightarrow$	girl   dog   bone
VP	$\rightarrow$	TV NP	ΤV	$\rightarrow$	bit
VP	$\rightarrow$	DV NP NP	DV	$\rightarrow$	gave

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#### What about incrementality?

First, what does it mean to be incremental?

- Each word is integrated into the parse as soon as it is seen/heard.
- Problems with current parse are detected immediately; also possible to make predictions about upcoming words.
- Evidence: eye-tracking of semantic interpretation, garden paths.

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- Human Parsing  $\bullet$  There are still 4 disconnected structures before the rule VP  $\rightarrow$  DV NP NP applies, reducing the number to 2 (after which, 1).



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#### Another problematic example

Consider the garden path sentence the old man the boats.

- Assume a serial bottom-up parser (or limited parallel-key is that the correct structure is not considered initially).
- At what point (intuitively) does a human realize the initial analysis is incorrect?
- At what point does the parser realize this?

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#### Another problematic example

Grammar for processing the old man the boats:

S	$\rightarrow$	NP VP	Det	$\rightarrow$	the   a
NP	$\rightarrow$	Det CN	Adj	$\rightarrow$	old
NP	$\rightarrow$	Det Adj CN	CN	$\rightarrow$	man   boats   old
VP	$\rightarrow$	TV NP	TV	$\rightarrow$	man   like
VP	$\rightarrow$	DV NP NP	DV	$\rightarrow$	gave

#### Garden path is too late!

The bottom-up parser doesn't realize its mistake until it reaches the end of the sentence, and cannot create a full parse:



But humans recognize a problem at the second the: they have an **expectation** about what should come next, and it is violated.

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#### Left Corner Parsing

**Left corner parsing** is more cognitively plausible: each word is immediately integrated into a single evolving structure which makes predictions about what will come next.

- Mixed directionality: constrained by input (like bottom-up) but also making predictions (like top-down).
- Chart contains active edges: incomplete constituents representing predictions.
- Ex: NP/CN is an incomplete constituent that will become a complete NP if a CN is seen next (cf. categorial grammar).



#### Example LC parse for garden path sentence

To try on your own with the grammar provided earlier: the old man the boats

Confirm that the parser realizes a problem where it should!

#### Summary so far

- RD parsing cannot model humans because of problems with (eg.) left recursion.
- SR parsing cannot model humans because it doesn't recognize garden paths immediately.
- CKY parsing cannot model humans because it is too parallel, or (if limited), also doesn't recognize garden paths immediately.
- So, where does that leave us?

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#### **Rules of Left Corner Parsing**

- 1. **Projection**: For a completed edge Y and a grammar rule  $X \rightarrow Y Z$ , add an active edge X/Z, where Y and X/Z span the same part of the string.
- 2. **Completion**: For an active edge X/Y and a completed edge Y that are adjacent, add a completed edge X that spans the width of both.
- 3. Composition: For two adjacent active edges X/Y and Y/Z, add an active edge X/Z that spans the width of both.

Rule 3 is not necessary for LC parsing, but is necessary for a fully incremental version (i.e., to ensure a single connected structure).

### Human Parsing Rules of Left Corner Parsing

If dealing with non-binary grammar:

- 1. **Projection**: For a completed edge Y and a grammar rule  $X \rightarrow Y \alpha$ , add an active edge  $X/\alpha$ , where Y and  $X/\alpha$  span the same part of the string.
- 2. Partial Completion: For an active edge X/Y  $\alpha$  and a completed edge Y that are adjacent, add an active edge X/ $\alpha$  that spans the width of both.
- 3. **Completion**: For an active edge X/Y and a completed edge Y that are adjacent, add a completed edge X that spans the width of both.
- 4. Composition: For two adjacent active edges X/Y and Y/Z, add an active edge X/Z that spans the width of both.

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#### Summary

- Left-corner parsing achieves full incrementality by using chart entries to represent partial/predictive syntactic structure.
- Looks promising for modelling ambiguity resolution and garden paths.
- But still haven't explained why some parses are preferred or some locally ambiguous sentences (but not others) cause garden paths.
- Other open research issues:
  - Developing fully incremental parsers for wide range of grammar formalisms (some easier than others).
  - How/when does semantics fit in?