COMP5046: Parsing with Grammars

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Parsing: determining nested sentence structure

- Structure
 - Why?
 - Applications
 - Grammars and formal languages
 - Context Free Grammars (CFGs)
 - Parsing algorithms
- 2 Statistics
 - Probabilistic Context Free Grammars (PCFGs)
 - Improving PCFG parsing
 - Dependency grammar
 - Dependency parsing
- Semantics
 - Semantic Role Labelling

Part I

Introduction to Syntax, Ambiguity and Parsing

Why bother parsing?

- Why is John ate the apple
- different to The apple ate John
- and different to The apple John ate
- and different to ate the John apple

Why bother parsing?

- Why is John ate the apple
- different to The apple ate John
- and different to The apple John ate
- and different to ate the John apple
- why doesn't the ate John apple make sense?

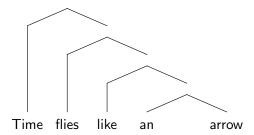


Why bother parsing?

- Why is John ate the apple
- different to The apple ate John
- and different to The apple John ate
- and different to ate the John apple
- why doesn't the ate John apple make sense?
- because there is more to language than just words!
- language has structure



Time flies like an arrow (Time (flies (like (an arrow))))



- words are grouped into constituents
- each constituent dominates the constituents under it
- the entire tree is a derivation



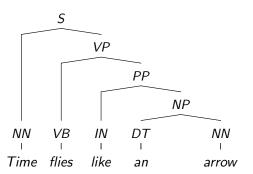
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Language structure influences our interpretation

Time flies like an arrow



Time moves forwards quickly, irreversible etc



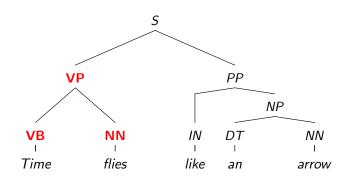
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Language structure *influences* our interpretation

Time flies like an arrow



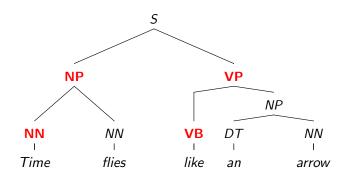
Time flies like you would time an arrow



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Language structure influences our interpretation

Time flies like an arrow



cf. Fruit flies like a banana



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Language structure influences our interpretation

Time flies like an arrow

but structure isn't all there is to it:

colourless green ideas sleep furiously

- some parses are more likely than others
- we are not dealing with the semantic meaning



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Applications

- practically every task involving grammatical text...
 - language modelling (and speech recognition)
 - word sense disambiguation
 - information extraction

 - question answering
 - automatic summarisation
 - (and making a comeback): machine translation
- doesn't work so well in noisy domains (the web?)
- a lot has to do with the training data available



- we don't use parsers when speed is important
- phrase structure parsing in $O(n^3)$ time (with large constants)
- parsers really grind to a halt on long sentences

Sometimes we can't use the detail

- other tasks we don't know how to use that much info (yet)
- many of these tasks are currently treated as bag of words:
 - text categorisation
 - information retrieval

(Formal) Languages

- formal languages are sets of strings
- these sets can potentially be infinite in size
 - e.g. all words made of the letter a
 - a, aa, aaa, . . .
- we want a way of compactly describing these sets
- we can describe methods of recognising (accepting) members
- these methods have different computational power (expressiveness)
- languages are grouped into language classes



Expressiveness of Language Classes

- how powerful is a language class?
- Chomsky hierarchy of language classes

LANGUAGE	AUTOMATA	
Regular	finite state	less complex
Context free	push-down	
Context sensitive	linear bounded	
Recursively enumerable	Turing machine	more complex

- where do natural (human) languages fit?
- what about programming languages?

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Regular Expressions

- simple yet powerful language
 - tokenisation
 - morphological analysis
- can be accepted by Finite State Automata/Transducers
- what can it represent?
 - language of words made of the letter a
 - a*
- what can't it represent?
 - recursion balanced brackets
 - Ø, (), ()(), (()), ...
 - expressions of the form A^nB^n
 - ab, aabb, aaabbb, ...

How do I parse HTML/XML with a regular expression?

You can't parse [X]HTML with regex. Because HTML can't be parsed by regex. Regex is not a tool that can be used to correctly parse HTML. As I have answered in HTML-and-regex questions here so many times before, the use of regex will not allow you to consume HTML. Regular expressions are a tool that is insufficiently sophisticated to understand the constructs employed by HTML. HTML is not a regular language and hence cannot be parsed by regular expressions. Regex queries are not equipped to break down HTML into its meaningful parts, so many times but it is not getting to me. Even enhanced irregular regular expressions as used by Perl are not up to the task of parsing HTML. You will never make me crack. HTML is a language of sufficient complexity that it cannot be parsed by regular expressions. Even Jon Skeet cannot parse HTML using regular expressions. Every time you attempt to parse HTML with regular expressions, the unholy child weeps the blood of virgins, and Russian hackers pwn your webapp. Parsing HTML with regex summons tainted souls into the realm of the living, HTML and regex go together like love, marriage, and ritual infanticide. The <center> cannot hold it is too late. The force of regex and HTML together in the same conceptual space will destroy your mind like so much watery putty. If you parse HTML with regex you are giving in to Them and their blasphemous ways which doom us all to inhuman toil for the One whose Name cannot be expressed in the Basic Multilingual Plane, he comes. HTML-plus-regexp will liquify the nerves of the sentient whilst you observe, your psyche withering in the onslaught of horror. Regex-based HTML parsers are the cancer that is killing StackOverflow it is too late it is too late we cannot be saved the trangession of a chilld ensures regex will consume all living tissue (except for HTML which it cannot, as previously prophesied) dear lord help us how can anyone survive this scourge using regex to parse HTML has doomed humanity to an eternity of dread torture and security holes using regex as a tool to process HTML establishes a breach between this world and the dread realm of corrupt entities (like SGML entities, but more corrupt) a mere glimpse of the world of regex parsers for HTML will instantly transport a programmer's consciousness into a world of ceaseless screaming, he comes, the pestilent slithy regex-infection will devour your HTML parser, application and existence for all time like Visual Basic only worse he comes he comes do not fight he comes, his unholy radiance destro ying all enlightenment, HTML tags leaking from your eyes fike liquid pain, the song of regular expression parsing will extinguish the voices of mortal man from the sp here I can see it can you see it it is beautiful the final snuf fing of the lies of Man ALL IS LOST ALL I S LOST the popy he comes he comes he comes the ichor permeates all MY FACE MY FACE or dod no NO NOOOO NO stop the an des ate not real ZALGO IS TONY THE PONY HE COMES

Have you tried using an XML parser instead?

http://stackoverflow.com/questions/1732348/

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Regular languages and natural language

- how do regular expressions apply to language?
- we know language has recursive structure
- regular languages can only represent bounded depth
- in terms of natural languages it means you can't represent:
 - \bullet NP \longrightarrow NP PP
 - I see [the girl] [on the hill] [with the telescope] [under the stars] [in the morning]...



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Formal representation for Context Free Grammars

- a set of *terminal* symbols: $V = \{w_1, w_2, \dots, w_n\}$ w_i is a word in the vocabulary (or alphabet)
- a set of non-terminal symbols: $N = \{N_1, N_2, \dots, N_m\}$
- a start symbol $N_s \subseteq N$ (often written as S in a grammar)
- rules of the form $N_i \longrightarrow \zeta_j$ where ζ_j is a sequence of terminals and nonterminals
- each rule application is dependent on a single symbol
- i.e. no rules of the form $XY \longrightarrow Z$



CFG examples

- language of words made of the letter a
 - terminal symbols: {a}
 - non-terminal symbols: $\{S\}$
 - start symbol: S
 - rules:
 - $S \longrightarrow Sa$
 - $S \longrightarrow \epsilon$
- balanced brackets
 - terminal symbols: {(,)}
 - non-terminal symbols: $\{S\}$
 - start symbol: S
 - rules:
 - $S \longrightarrow (S)S$
 - $S \longrightarrow \epsilon$

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Context Free Grammars

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- representation of recursive structures
- can deal with balancing brackets problem
- cannot deal with expressions of the form $A^nB^nC^n$
- allows for most constructions in many natural languages (e.g. English)
- can be represented/executed by a pushdown automaton



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An example grammar

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```
S \rightarrow NP VP
                                      |Det \rightarrow that | this | a
                                       Noun \rightarrow book \mid flight \mid meal \mid money
S \rightarrow Aux NP VP
                                        Verb \rightarrow book \mid include \mid prefer
S \rightarrow VP
NP \rightarrow Det\ Nominal
                                       Aux \rightarrow does
Nominal \rightarrow Noun
Nominal \rightarrow Noun \ Nominal \ \| Prep \rightarrow from \ | \ to \ | \ on
                                        Proper-Noun → Houston | TWA
NP \rightarrow Proper-Noun
VP \rightarrow Verb
VP \rightarrow Verb NP
                                       Nominal \rightarrow Nominal PP
```

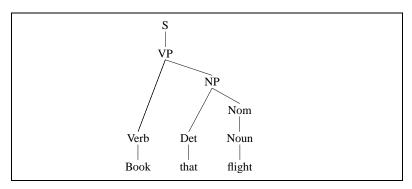
Jurafsky and Martin, Figure 10.2



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An example parse

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Jurafsky and Martin, Figure 10.1

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Context Sensitive Grammars

- can use surrounding categories to determine rule application $\zeta_i \longrightarrow \zeta_i$
- needed to handle some constructions in natural language (e.g. crossing dependencies in Dutch)
- · cost of parsing with these grammars is painful
- try to minimise context dependence use mildly context sensitive grammars
- otherwise extremely expensive to parse



Part II

Parsing Algorithms

Parsing = search

- grammars are purely declarative
 - they don't say how to find a derivation
 - they don't say how to choose the best derivation
- the grammar defines the search space
- the constraints are the sentence and the grammar



Parsing Algorithms

- top down parsing
 - try to build the parse tree from the root downwards
 - hypothese all trees (smallest first)
- bottom up parsing
 - try to build the tree from the leaves (words) upwards
 - hypothesise all subtrees from the sentence
- Earley algorithm (top down + memoization)
- CKY Cocke Younger Kasami (bottom up + memoization)



Top-down parsing

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- try to build the parse tree from the root downwards
- hypothesising all trees (smallest first)



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Here's the grammar again

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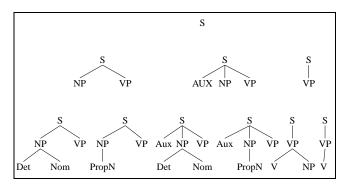
```
S 	oup NP VP
S 	oup Aux NP VP
S 	oup VP
NP 	oup Det Nominal
Nominal 	oup Noun
Nominal 	oup Noun Nominal
NP 	oup Proper-Noun
NP 	oup Verb
NP 	oup Verb
NP 	oup Verb
NP 	oup Verb
NP 	oup NP
NOminal 	oup Noun Nominal
NOminal 	oup Noun Nominal
NOminal 	oup Nominal 	oup Nominal
NOminal 	oup Nominal 	oup
```

Jurafsky and Martin, Figure 10.2



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Top-down example



Jurafsky and Martin, Figure 10.3

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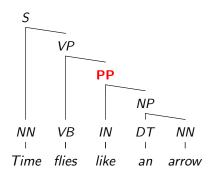
Problems with top-down parsing

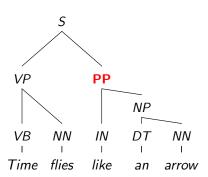
- search pays (almost) no attention to the input
- so we generate many many options which aren't possible
- ambiguity causes parts of the tree to be rebuilt
- this duplication is massive in useful grammars



Duplication

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Bottom-up parsing

- try to build the tree from the leaves (words) upwards
- hypothesise all subtrees from the sentence



Here's the grammar again

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```
S 	o NP \ VP
S 	o Aux \ NP \ VP
S 	o VP
NP 	o Det \ Nominal
Nominal 	o Noun
Nominal 	o Noun
Nominal 	o Noun \ Nominal
NP 	o Proper-Noun
VP 	o Verb
VP 	o Verb \ NP

Net 	o Det 	o that | this | a
Noun 	o book | flight | meal | money
Noun 	o book | include | prefer
Aux 	o does

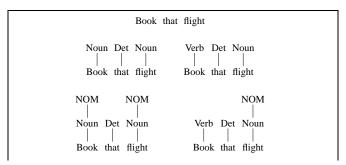
Noun 	o book | include | prefer
Aux 	o does
Nominal 	o Noun 	o Houston | TWA
Nominal 	o Nominal 	o Nominal \ PP
```

Jurafsky and Martin, Figure 10.2



Bottom-up example

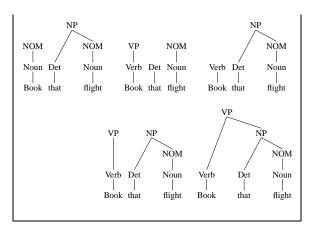
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Jurafsky and Martin, Figure 10.4

Bottom-up example

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Jurafsky and Martin, Figure 10.4 (cont)



Problems with bottom-up parsing

- search pays no attention to the target (the root)
- generate many options which can/will never be used
- ambiguity causes parts of the tree to be rebuilt
- this duplication is massive in useful grammars



Top-down depth-first parsing

- in practice the space is not explored in parallel
- depth first allows you to reach the words themselves as soon as possible
 - this means you can throw away incorrect parses earlier
- when a tree no longer matches backtrack
- recursive descent parsing



Left-branching problem

- when a grammar rule is of the form: $X \longrightarrow X Y$
- for example prepositional phrases in NPs: $NP \longrightarrow NP PP$
- have to factor the rule into multiple rules
- these rules are no longer natural

Memoisation

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- both top-down and bottom-up rebuild trees (over and over)
- memoisation involves saving previous calculations
- top-down + memoisation = Earley algorithm
- bottom-up + memoisation = CKY algorithm
- chart stores the cached calculations



Charts

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- represent/store the constructed subtrees
- later we will see that probabilities are stored too
- handy visualisation of alternative parses (see NLTK)
- spans are represented as two indices (beginning and end)
- the indices fall between the words starting from zero
- words are the first spans added

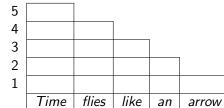


Cocke Kasami Younger (CKY) algorithm

- build all possible subtrees from the leaves
- store subtrees in the chart
- can build bottom to top or left to right
- time complexity is $O(n^3)$ in the sentence length
- we will examine the probabilistic case later

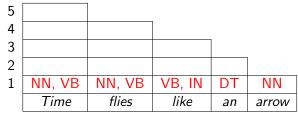






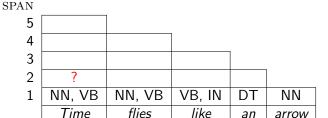
CKY algorithm example





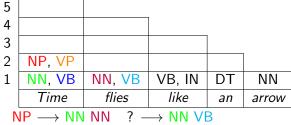
Fill in the bottom level POS tags

CKY algorithm example



Start moving up the chart, one cell at a time

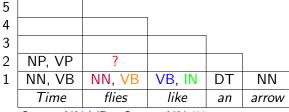
SPAN



 $VP \longrightarrow VB NN ? \longrightarrow VB VB$



SPAN

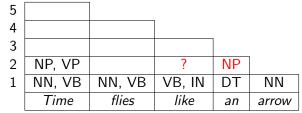


- ? \longrightarrow NN VB ? \longrightarrow NN IN
- ? \longrightarrow VB VB ? \longrightarrow VB IN

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CKY algorithm example





Filling in the rest of the row

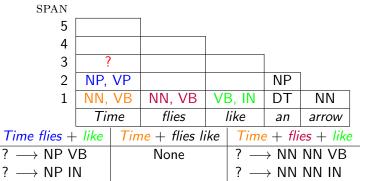
SYDNEY

? \longrightarrow NN VB IN

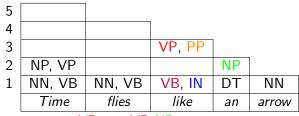
CKY algorithm example

? \longrightarrow VP VB

? \longrightarrow VP IN



SPAN

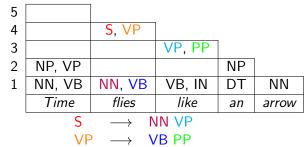


$$VP \longrightarrow VB NP$$

$$PP \longrightarrow IN NP$$

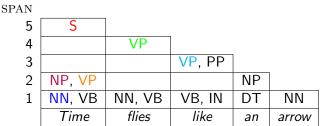
CKY algorithm example

SPAN



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 $S \longrightarrow NN VP$

Time moves forwards quickly, irreversible etc

 \rightarrow NP VP | cf. Fr.

cf. Fruit flies like a banana

 $S \rightarrow VPPP$

Time flies like you would time an arrow

Extra complications

- we have a recogniser, not a parser
 - need to keep back-pointers to remember the whole tree
- unary rules, need to be applied to every cell
- rules are often binarised to reduce the number of potential derivations
 - this also reduces the effectiveness of the rule probabilities
- where does the grammar come from?
 - hand-written
 - data-driven
- how do you choose between the final parse trees?



Take away

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- Natural language is not regular
- What a CFG looks like
- Over-generation, under-generation
 - like regular expressions
- Syntactic ambiguity is everywhere
 - most often: preposition attachment, coordination, noun phrase
- PCFGs: probability of sentence = product of probability of production rules
- Dynamic programming gives $O(n^3)$ parse
 - may decide ambiguities

