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Pre-processing and annotation	
Raw data from a linguistic source can't be exploited directly. We firs to perform:	t have
 <i>pre-processing:</i> identify the basic units in the corpus: – tokenization; – sentence boundary detection; 	
 <i>annotation:</i> add task-specific information: parts of speech; syntactic structure; dialogue structure, prosody, etc. 	
– thatogue suleture, prosody, etc.	
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Tokenization	
<i>Tokenization:</i> divide the raw textual data into tokens (words, number punctuation marks).	rs,
<i>Word:</i> a continuous string of alphanumeric characters delineated by whitespace (space, tab, newline).	
Example: potentially difficult cases:	
• amazon.com, Micro\$oft	
• John's, isn't, rock'n'roll	
 child-as-required-yuppie-possession (As in: "The idea of a child-as-required-yuppie-possession must motivating them.") 	t be
• cul de sac	
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Corpus Annotation	
<i>Annotation:</i> adds information that is not explicit in the datits usefulness (often application-specific).	ta itself, increases
<i>Annotation scheme:</i> basis for annotation, consists of a tag annotation guidelines.	set and
Tag set: is an inventory of labels for markup.	
<i>Annotation guidelines:</i> tell annotators (domain experts) he applied; ensure consistency across different annotators.	ow tag set is to be
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Part-of-speech (POS) annotation	
Part-of-speech (POS) tagging is the most basic kind of linguistic anno	otation.
Each linguistic token is assigned a code indicating its <i>part of speech</i> , basic grammatical status.	i.e.,
Examples of POS information:	
• singular common noun;	
• comparative adjective;	
• past participle.	
POS tagging forms a basic first step in the disambiguation of homogr	aphs.
E.g., it distinguishes between the verb "boot" and the noun "boot".	
But it does not distiguish between "boot" meaning "kick" and "boot" "boot a computer", both of which are transitive verbs.	as in
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Inf1-DA 2010-2011 II: 88 / 119 Example POS tag sets • CLAWS tag set (used for BNC): 62 tags; (Constituent Likelihood Automatic Word-tagging System) • Brown tag set (used for Brown corpus): 87 tags: • Penn tag set (used for the Penn Treebank): 45 tags. Category CLAWS Examples Brown Penn Adjective happy, bad AJ0 JJ JJ Adverb often, badly PNI $\mathbf{C}\mathbf{D}$ CD DT DT Determiner this, each DT0 Noun aircraft, data NN0 NN NN Noun singular woman, book NN1 NN NN Noun plural NN2 NN NN women, books NP NNP Noun proper singular London, Michael NP0 Noun proper plural Australians, NP0 NPS NNPS Methodists

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POS Tagging	
Idea: Automate POS tagging: look up the PO	DS of a word in a dictionary.
Problem: POS ambiguity: words can have se	everal possible POS's; e.g.:
Time flies like an arrow.	(1)
time: singular noun or a verb;	
flies: plural noun or a verb;	
like: singular noun, verb, preposition.	
Combinatorial explosion: (1) can be assigned	d $2 imes 2 imes 3 = 12$ different
POS sequences.	
Need more information to resolve such ambig	guities.
It might seem that higher-level meaning (sem	antics) would be needed, but in
fact great improvement is possible using the	probabilities of different POS.
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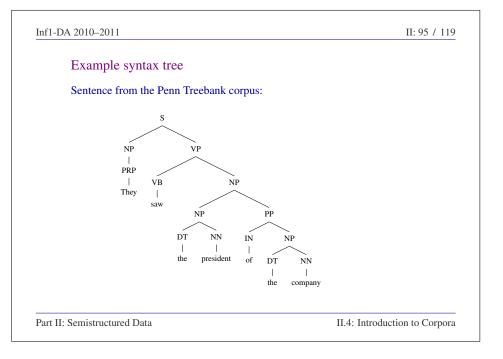
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Probabilistic POS tagging	
<i>Observation:</i> words can have more than one POS, but one of frequent than the others.	of them is more
<i>Idea:</i> assign each word its most frequent POS (get frequent manually annotated training data). Accuracy: around 90%.	
<i>Improvement:</i> use frequencies of POS sequences, and other Accuracy: 96–98%.	r context clues.
Example output from a POS tagger (not XML format!):	
Our/PRP\$ enemies/NNS are/VBP innovative/JJ and/C0 resourceful/JJ ,/, and/CC so/RB are/VB we/PRP ./. The never/RB stop/VB thinking/VBG about/IN new/JJ way to/TO harm/VB our/PRP\$ country/NN and/CC our/PR	ey/PRP ys/NNS
people/NN, and/CC neither/DT do/VB we/PRP ./.	(George W. Bush)
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Use of markup languages	
An important general application of markup languag separate data from metadata.	ges, such as XML, is to
In a corpus, this serves to keep different types of inf	formation apart;
• <i>Data</i> is just the raw data. In a corpus this is the text itself.	
• <i>Metadata</i> is data about the data. In a corpus this is the various annotations.	
Nowadays, XML is the most widely used markup la	inguage for corpora.
The example on the next slide is taken from the BN was released only in 2007.	C XML Edition, which
(The previous BNC World Edition was formatted in	SGML.)
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Example from the BNC XML Edition	
<pre><wtext type="FICTION"></wtext></pre>	
<div level="1"></div>	
<head> <s n="1"></s></head>	
<w c5="NN1" hw="chapter" pos="SUBST">CHAPTER </w>	
<w c5="CRD" hw="1" pos="ADJ">1</w>	
<s n="2"></s>	
<c c5="PUQ"> </c>	
<w c5="CJC" hw="but" pos="CONJ">But</w>	
<c c5="PUN">,</c> <c c5="PUQ"> </c>	
<w c5="VVD" hw="say" pos="VERB">said </w>	
<w c5="NP0" hw="owen" pos="SUBST">Owen</w>	
<c c5="PUN">,</c> <c c5="PUQ"> </c>	
<w c5="AVQ" hw="where" pos="ADV">where </w>	
<w c5="VBZ" hw="be" pos="VERB">is </w>	
<w c5="AT0" hw="the" pos="ART">the </w>	
<w c5="NN1" hw="body" pos="SUBST">body</w>	
<c c5="PUN">?</c> <c c5="PUQ"> </c>	
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Aspects of this example	
This example is the opening text of J10, a n	ovel by Michael Pearce.
Some aspects of the tagging:	
• The wtext element stands for <i>written</i> indicates the genre.	text. The attribute type
• The head element tags a portion of he heading).	ader text (in this case a chapter
• The s element tags sentences. (N.B., a sentence.) Sentences are numbered via	
• The w element tags words. The attribut detailed POS information given by the CLAWS code. The attribute hw repress (e.g., the root form of "said" is "say").	c5 attribute, which contains the
• The c element tags punctuation.	
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Syntactic annotation (parsing)	
Syntactic annotation: information about the	structure of sentences.
Prerequisite for computing meaning.	
Linguists use phrase markers to indicates wh together:	tich parts of a sentence belong
• noun phrase (NP): noun and its adjective	es, determiners, etc.
• verb phrase (VP): verb and its objects;	
• prepositional phrase (PP): preposition and	nd its NP;
• sentence (S): VP and its subject.	
Phrase markers group hierarchically in a syn	tax tree.
Syntactic annotation can be automated. Accu	uracy: around 90%.
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The same syntax tree in XML:	
<s></s>	
<np><w pos="PRP">They</w></np>	
<vp><w pos="VB">saw</w></vp>	
<np></np>	
<np><w pos="DT">the</w> <w pos="DT">the</w></np>	os="NN">president
<pp><w pos="NN">of</w></pp>	
<pre><np><w pos="DT">the</w> <w< pre=""></w<></np></pre>	pos="NN">company
Note the conventions used in the above do	ocument: phrase markers are
represented as elements; whereas POS tag	gs are given as attribute values.
N.B. The tree on the previous slide is <i>not</i>	the XML element tree generated
by this document.	