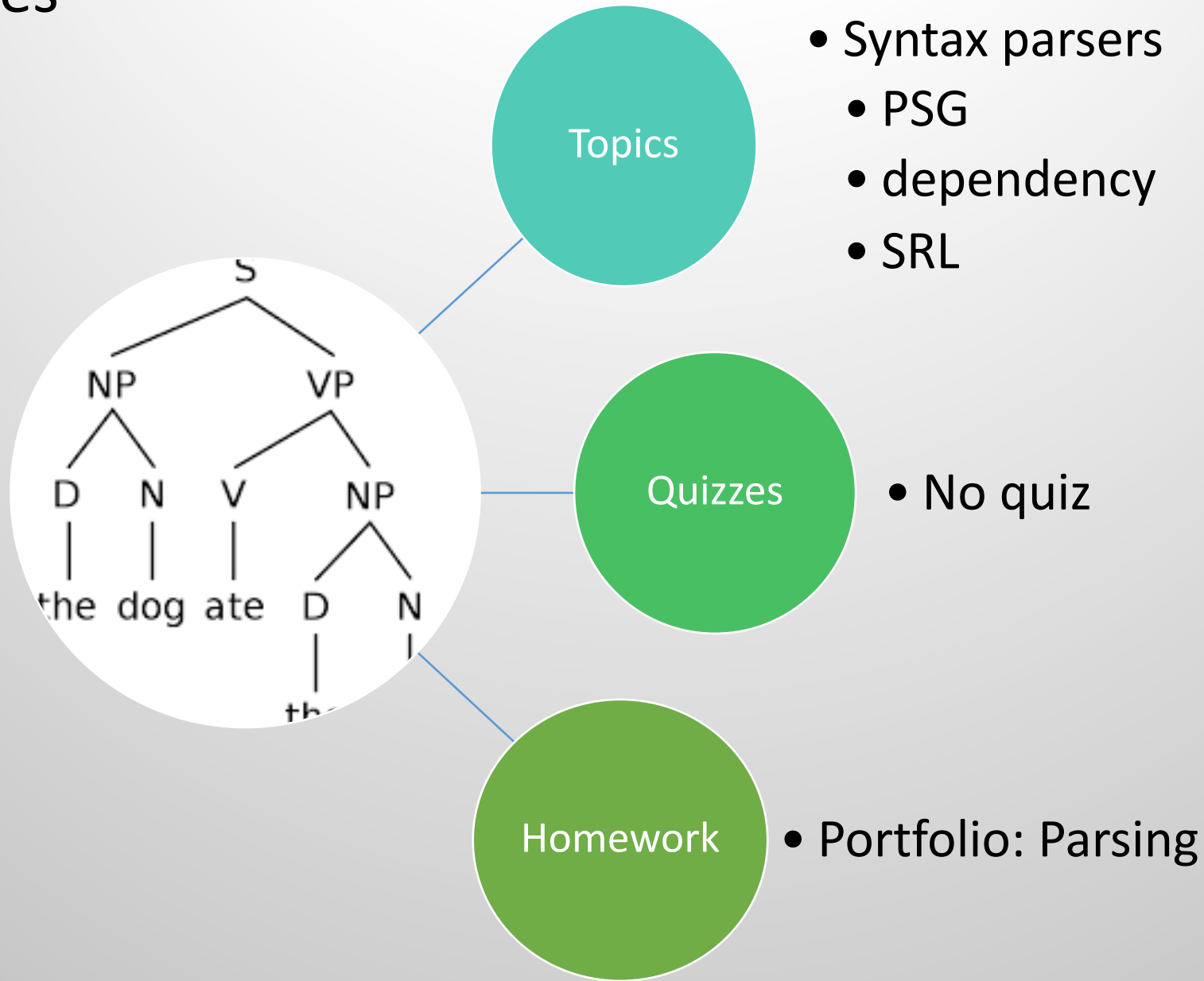


Natural Language Processing

Dr. Karen Mazidi



Part Three: Sentences



Parsing natural language

- Formal grammars rely on rules; natural language requires a probabilistic approach
- Issues:
 - POS ambiguity, example:
 - My spidey sense senses danger.
 - Structural ambiguity:

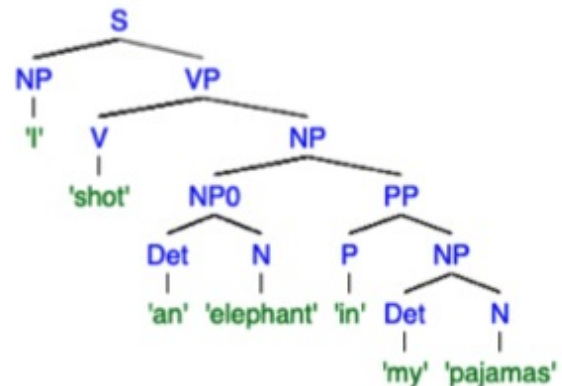


Figure 10.1: The elephant is in the pajamas

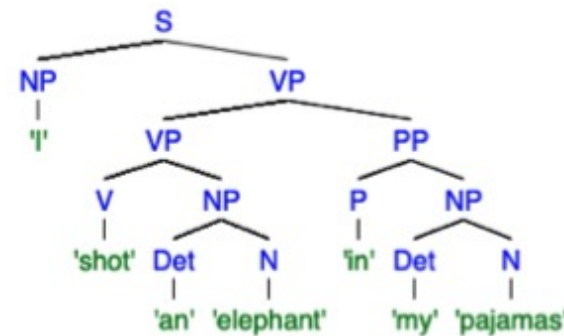


Figure 10.2: The shooter is in the pajamas

Structural ambiguity

- **Attachment ambiguity** – where to place a phrase in the syntax tree (elephant example)
- **Coordination ambiguity** – coordinating conjunctions
 - Old men and dogs

Types of syntax parsers

- Phrase structure grammar parse (PSG)
- Dependency parse
- Semantic role label parse

PSG parse

aka constituent parsing



PSG parse

- Organizes sentence in hierarchy of constituents (phrases)
 - Similar to CFG in that top level is S
 - Next-to bottom level is POS
 - Bottom level is the tokens
- PSG parsers are trained on millions of sentences
- Visualized in bracket notation or a tree
- Sometimes called **constituency parsing**

PSG

- <http://mshang.ca/syntree/>

```
[S [NP [DT The] [NN butler]]  
  [VP [VBN murdered]  
      [NP [NNP John]]  
      [PP [IN in] [DT the] [NP [NN living] [NN room]]]]
```

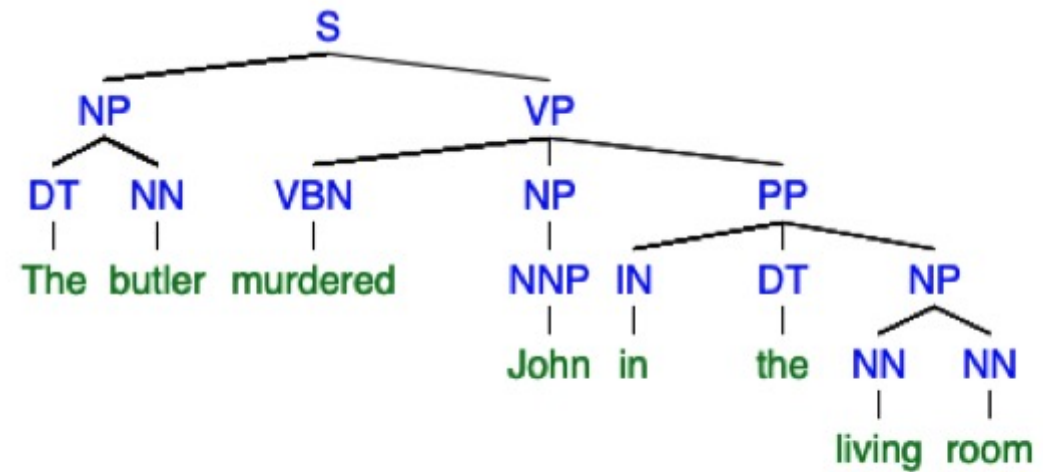


Figure 10.3: Phrase Structure Parse

AllenNLP

- Expanding boxes instead of tree
- <https://demo.allennlp.org/>



Figure 10.4: Top Level Phrase Structure

Expanding

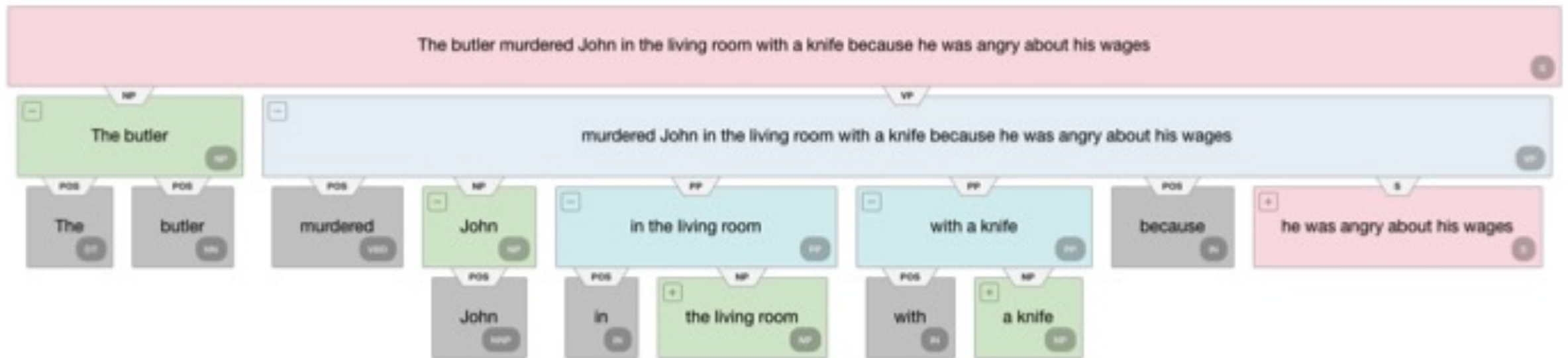


Figure 10.5: Adding Levels to the Phrase Structure

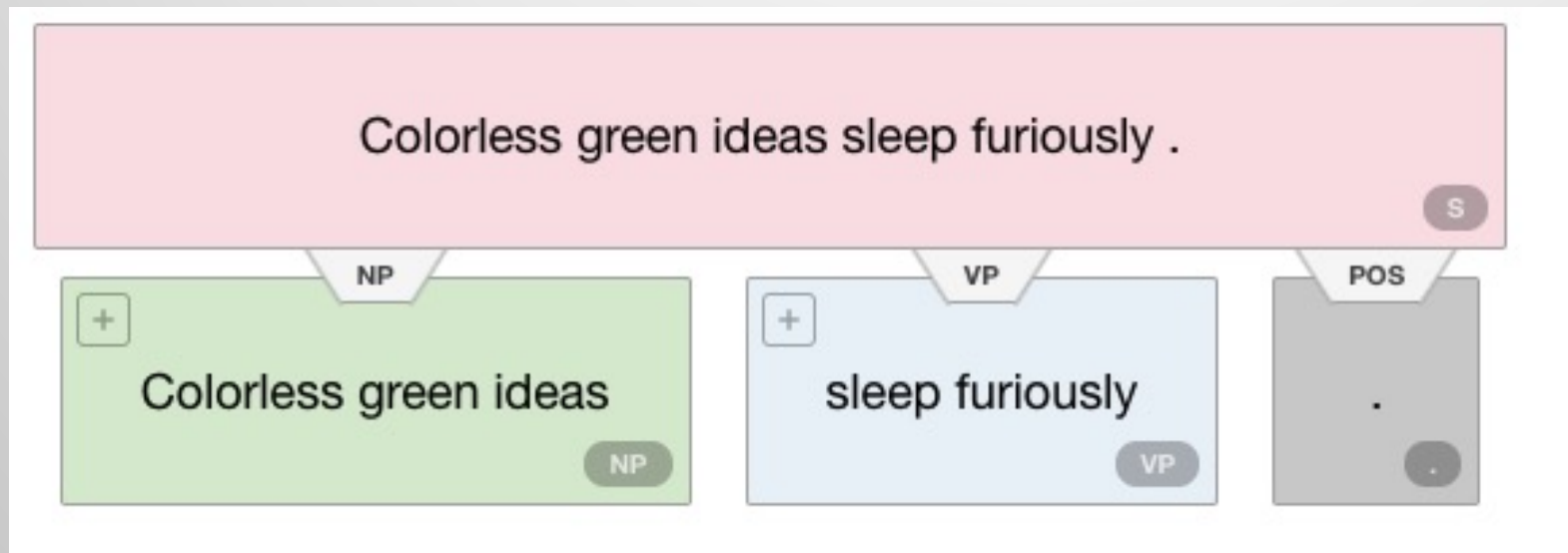
Expanding



Figure 10.6: Clause Structure

Example

- Colorless green ideas sleep furiously.



Dependency parse

DAG



Dependency parse

- Words are in an acyclic graph, usually with the main predicate as the root node
- Stanford dependencies in Figure 10.11
- Run at: <https://corenlp.run/>

Basic Dependencies:

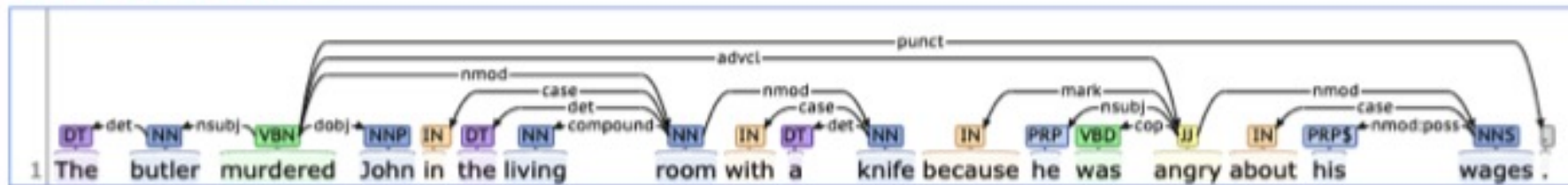
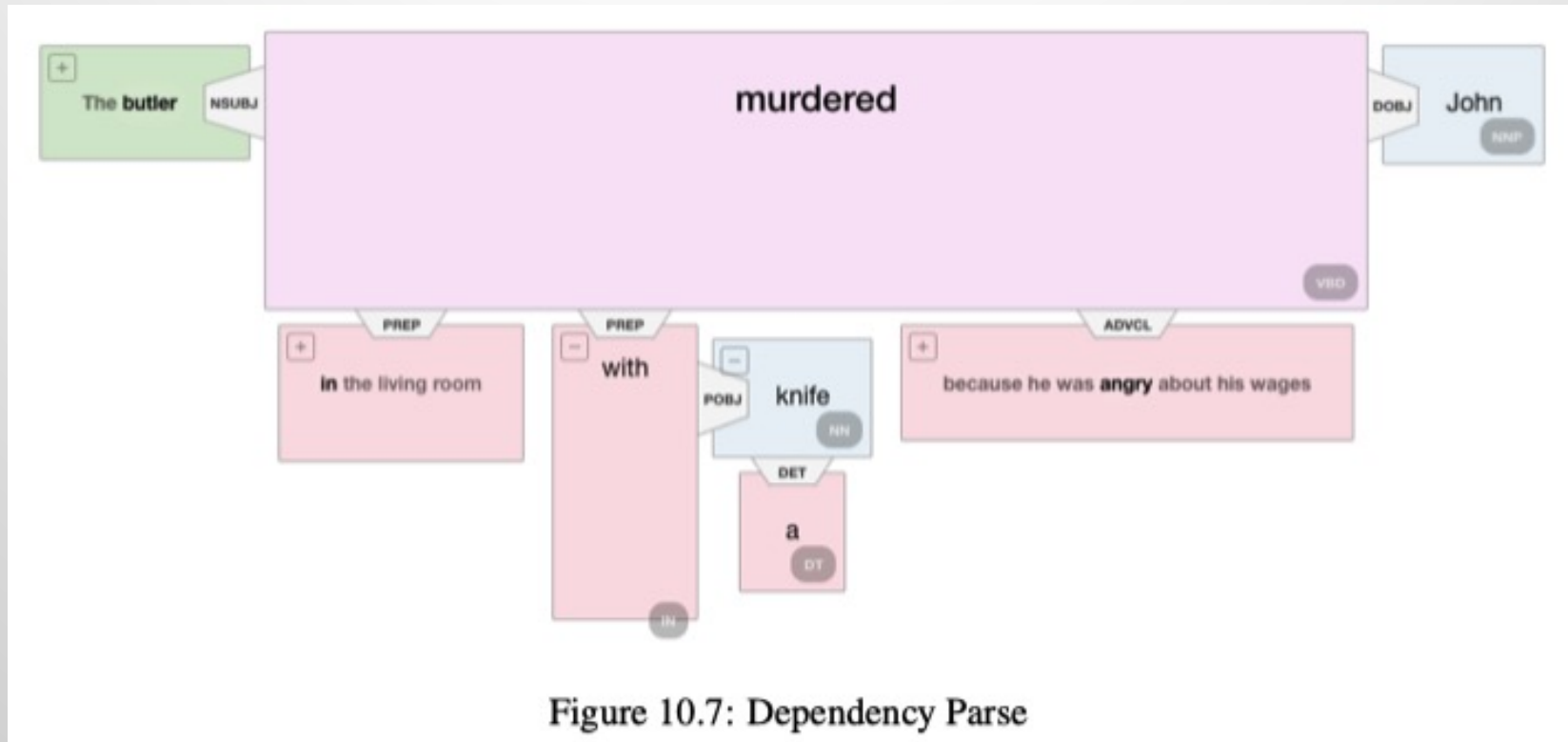


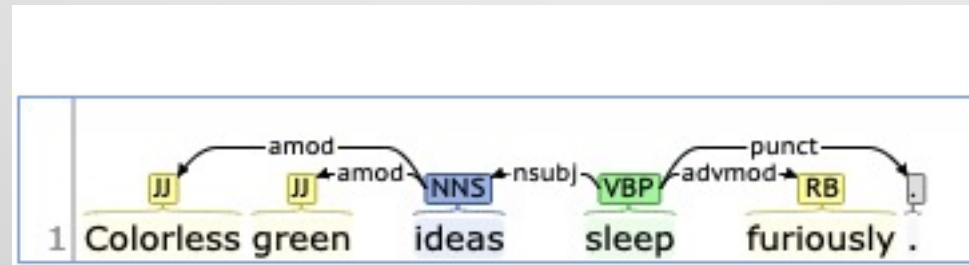
Figure 10.8: Dependency Parse, Another View

Allen NLP



Example

- Colorless green ideas sleep furiously.
- AllenNLP demo – problem for this sentence
- Stanford demo:



Copular clauses

- John is handsome.
 - Links subject with adjective
- John is a lawyer.
 - Links subject with noun

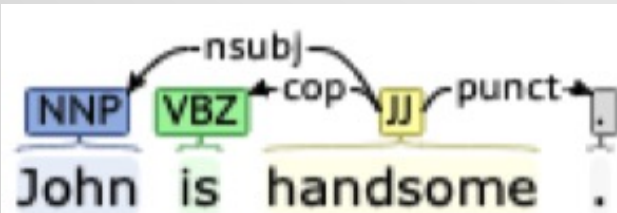


Figure 10.9: The predicate is JJ

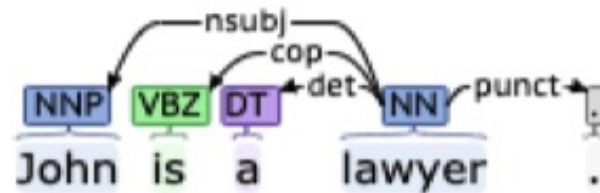


Figure 10.10: The predicate is NN

Copular clauses

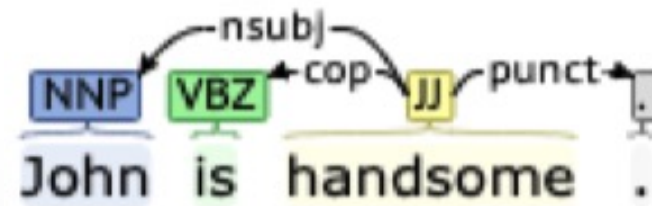


Figure 10.9: The predicate is JJ

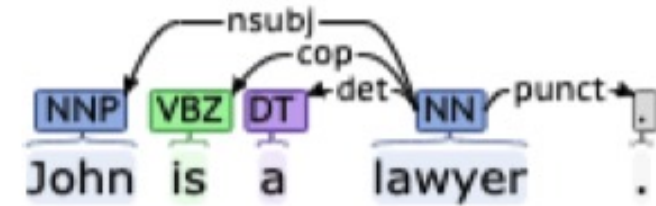


Figure 10.10: The predicate is NN

- Why isn't "is" the predicate?
- Evidence yes:
 - Verb "is" changes tense
 - Verb "is" matches subject
- Evidence no:
 - Move copular clause to subordinate position

Mary thinks that John is handsome.
Mary thinks John handsome.

Stanford dependencies

- See p. 118 of the Stanford manual
- Note the structure:
 - root
 - dep
 - aux
 - arg
 - agent
 - comp
 - subj

Universal dependencies

- Dependencies that work across languages
- <https://www.asc.ohio-state.edu/demarneffe.1/papers/depling.pdf>
- Stanford now uses universal dependencies by default:
- <https://nlp.stanford.edu/software/stanford-dependencies.html>

Common sentence patterns

- In expository text

Pattern	Example	Comment
S-V	The Cowboys lost.	no object or complement
S-V-dobj	The Cowboys lost the game.	direct object
S-V-iobj-dobj	The Cowboys handed them the game.	both direct and indirect objects
S-V-acomp	He looked tired.	'tired' completes the meaning
S-V-ccomp	I hope that you get the job.	'you' is the internal subject
S-V-xcomp	I love to take long walks.	no internal subject in clause
S-V-xcomp	I love walking in the woods.	no internal subject in clause

Table 10.1: Common Sentence Patterns

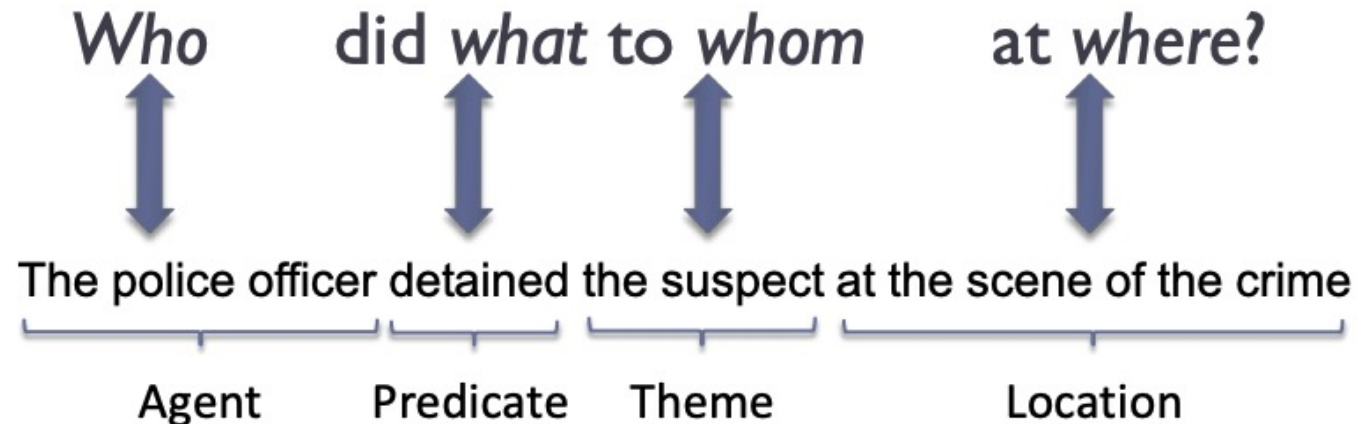
SRL parse

shallow semantics



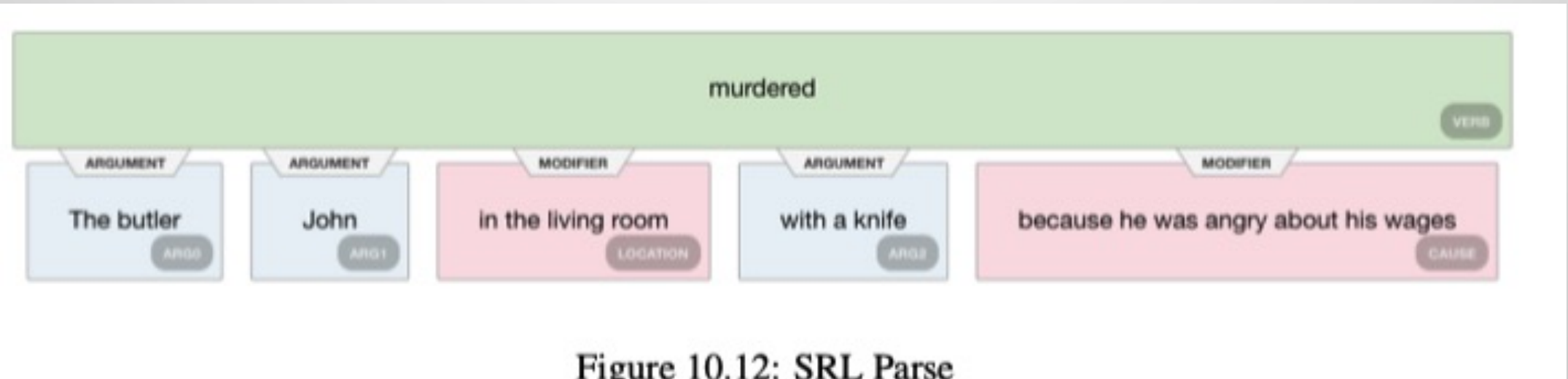
SRL Parse

- Semantic role label parse – determines a role for each constituent relative to the predicate
- Aka shallow semantic parse
- Two categories of labels:
 - Arguments: actors in the sentence
 - Modifiers: more details such as time and place



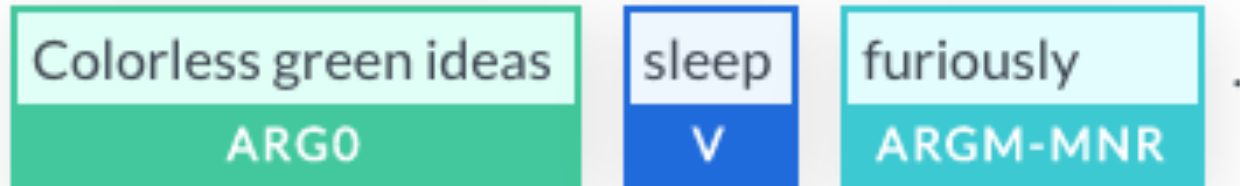
SRL

- AllenNLP



SRL

- AllenNLP



SRL arguments

- Numbered from 0 to 6
- Arguments available for a particular verb, particular sense of the word, vary
- Arg0, A0, is the agent of the sentence, the one doing the action
- Arg1, A1, is often the passive actor
- Arg1 vs. Arg0 has nothing to do with “subject”

```
Arg0 [The butler] murdered Arg1 [John].  
Arg1 [John] was murdered by Arg0 [the butler].
```

SRL arguments

- Arg2, A2, is often the ‘instrument’
 - With a knife
 - An argument can be a PP as well as a NP

Arg	Meaning	Example
Agent	Entity doing the action	John broke the window.
Patient	Entity that is acted upon	John broke the window .
Instrument	Entity used in action	John broke the window with a hammer .
Beneficiary	Entity recipient	John gave the ring to Mary .

Table 10.2: Common Arguments

SRL args

- More arguments

Thematic Role	Definition	Example
AGENT	The volitional causer of an event	<i>The waiter</i> spilled the soup.
EXPERIENCER	The experiencer of an event	<i>John</i> has a headache.
FORCE	The non-volitional causer of the event	<i>The wind</i> blows debris from the mall into our yards.
THEME	The participant most directly affected by an event	Only after Benjamin Franklin broke <i>the ice</i> ...
RESULT	The end product of an event	The city built a <i>regulation-size baseball diamond</i> ...
CONTENT	The proposition or content of a propositional event	Mona asked “ <i>You met Mary Ann at a supermarket?</i> ”
INSTRUMENT	An instrument used in an event	He poached catfish, stunning them <i>with a shocking device</i> ...
BENEFICIARY	The beneficiary of an event	Whenever Ann Callahan makes hotel reservations <i>for her boss</i> ...
SOURCE	The origin of the object of a transfer event	I flew in <i>from Boston</i> .
GOAL	The destination of an object of a transfer event	I drove <i>to Portland</i> .

- Problem: hard to create a standard set of arguments for all verbs
- PropBank: fewer roles

PropBank

- Palmer, Martha, Daniel Gildea, and Paul Kingsbury. 2005. The Proposition Bank: An Annotated Corpus of Semantic Roles. *Computational Linguistics*, 31(1):71–106
- Some roles:
 - Arg0: proto-agent
 - Arg1: proto-patient
 - Arg2: often benefactive, instrument, attribute, end state
 - Arg3: start point, benefactive, instrument, attribute
 - Arg4: the end point

SRL modifiers

- Not arguments, optional content that adds more info
- Common modifiers:

Mod	Meaning	Example
DIR	Motion along a path	John threw the papers in the trash .
LOC	Where the action happened	John was born in Texas .
MNR	How the action was performed	John broke the window violently .
TMP	When the action happened	John was born in 1960 .
CAU	Reason for action	John moved to NY because of his job .
PNC	Motivation for an action	John saved money for his move .

PropBank Verb Frames

- Annotation of major verbs and major senses
- Example:

fall.01

Arg1: Logical subject, patient, thing falling

Arg2: Extent, amount fallen

Arg3: start point

Arg4: end point, end state of arg1

Ex1: [Arg1 Sales] *fell* [Arg4 to \$25 million] [Arg3 from \$27 million].

Ex2: [Arg1 The average junk bond] *fell* [Arg2 by 4.2%].

PropBank Verb Frames

- Example verb frame:

increase.01 “go up incrementally”

Arg0: causer of increase

Arg1: thing increasing

Arg2: amount increased by, EXT, or MNR

Arg3: start point

Arg4: end point

- Clarifies commonalities here:

[Arg0 Big Fruit Co.] increased [Arg1 the price of bananas].

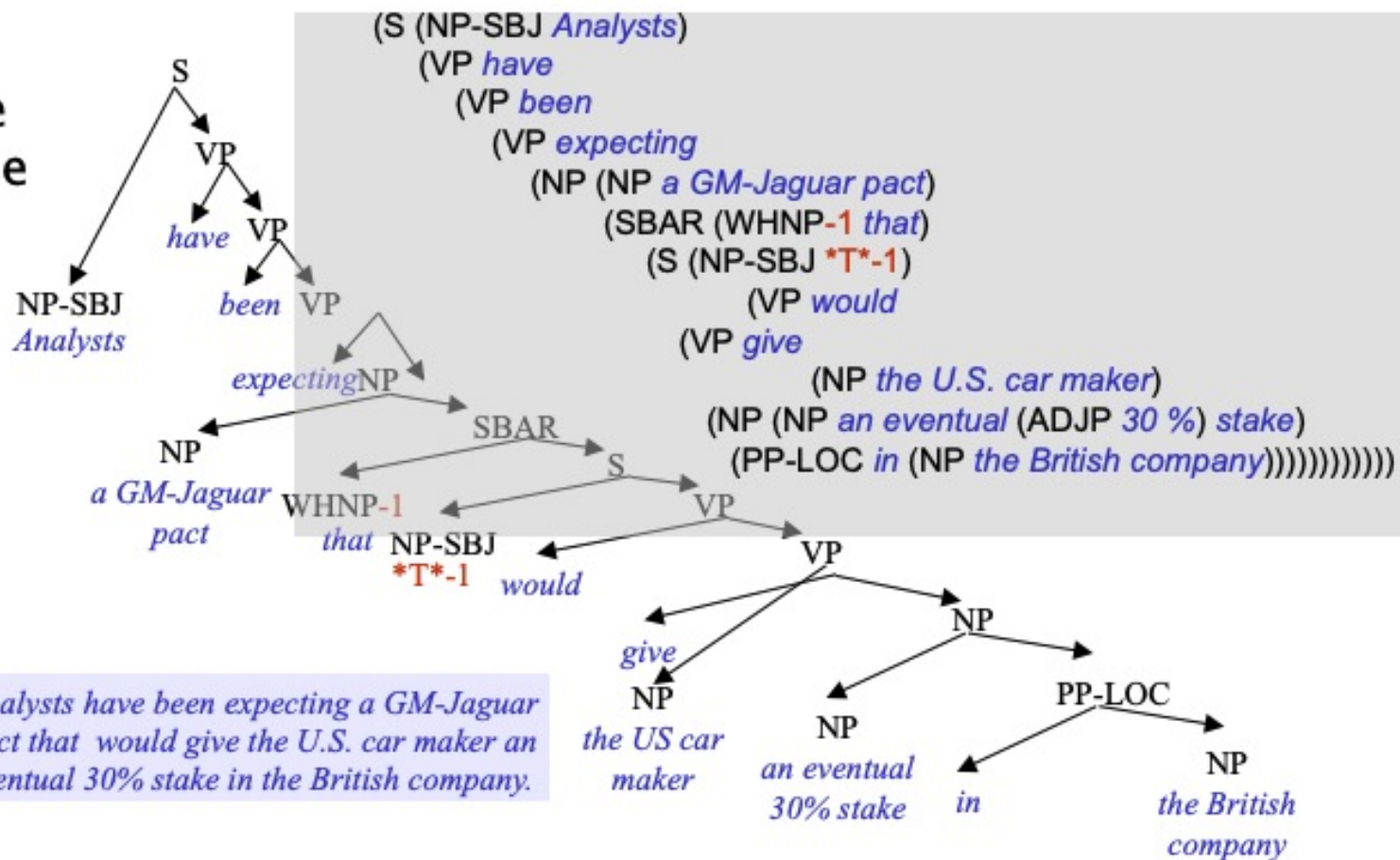
[Arg1 The price of bananas] was increased again [Arg0 by Big Fruit Co.]

[Arg1 The price of bananas] increased [Arg2 5%].

PropBanking a Sentence

Martha Palmer 2013

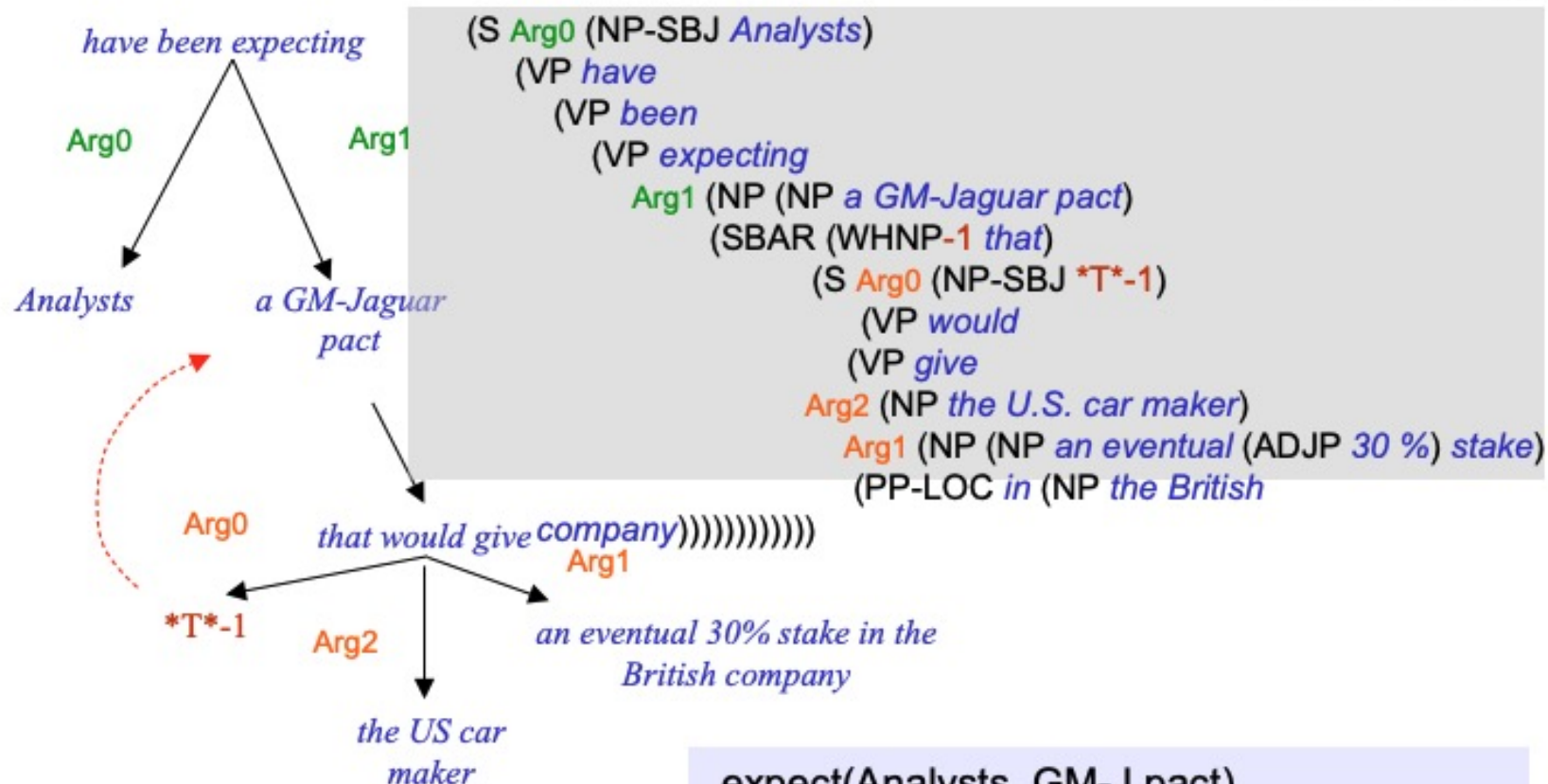
A sample
parse tree



Analysts have been expecting a GM-Jaguar pact that would give the U.S. car maker an eventual 30% stake in the British company.

The same parse tree PropBanked

Martha Palmer 2013



expect(Analysts, GM-J pact)
give(GM-J pact, US car maker, 30% stake)

SRL

- SRL gets closer to the semantics of the sentence
- Some have used SRL parses to extract logical propositions from text

Sasha broke the window

Pat opened the door

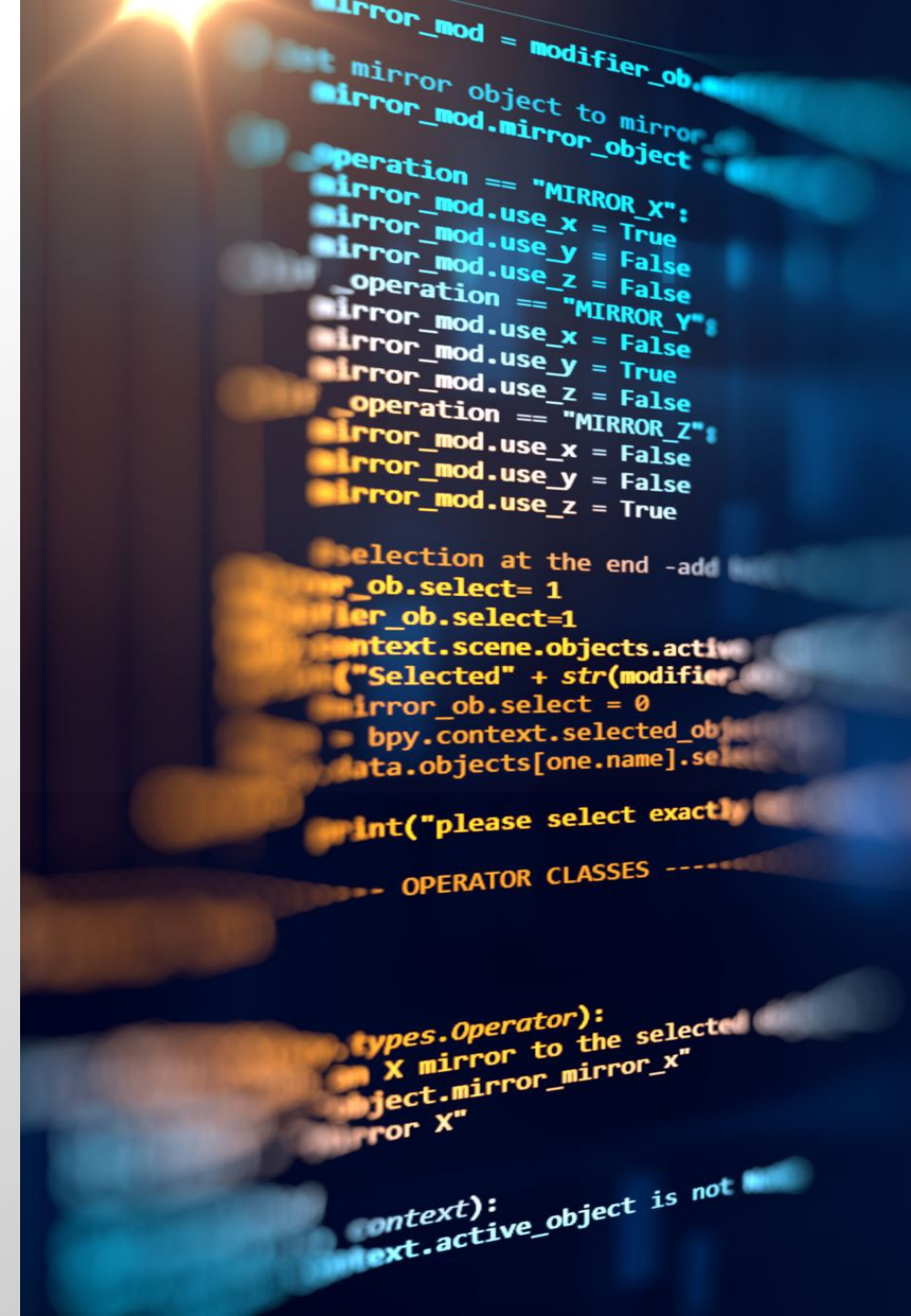
$$\exists e, x, y \text{ Breaking}(e) \wedge \text{Breaker}(e, \text{Sasha})$$
$$\wedge \text{BrokenThing}(e, y) \wedge \text{Window}(y)$$
$$\exists e, x, y \text{ Opening}(e) \wedge \text{Opener}(e, \text{Pat})$$
$$\wedge \text{OpenedThing}(e, y) \wedge \text{Door}(y)$$

SRL parsers

- Allen NLP
- SENNA

Other Software Tools

- Stanford
- spaCy



Stanford CoreNLP

- Toolkit from Stanford NLP group
- Many annotations, including tokenization, POS tagging, dependency parsing, and more
- To use Stanford CoreNLP:
 - 1. download CoreNLP
 - 2. Unzip the download
 - 3. Update classpath variables to point to the expanded directory

Stanford CoreNLP

- Setting up annotations in Java

Code 10.4.1 — **Stanford CoreNLP.** Java implementation

```
// Select the annotations
Properties props = new Properties();
props.setProperty("annotators", "tokenize, ssplit, pos, lemma, ner, parse,
                                dcoref, sentiment");

// Set up the pipeline
StanfordCoreNLP pipeline = new StanfordCoreNLP(props);

// Run all the selected Annotators on this text
pipeline.annotate(annotation);
```


Python Wrapper: Stanza

- install:

```
$pip3 install stanza  
$stanza.download('en')
```

Code 10.4.2 — Stanford CoreNLP. Stanza Python wrapper

```
import stanza  
  
# set up the pipeline  
nlp = stanza.Pipeline('en')  
  
# set up the doc object on text  
text = "Barack Obama was born in Hawaii. He was elected president in 2008."  
doc = nlp(text)  
  
for sentence in doc.sentences:  
    sentence.print_dependencies()
```

```
('Barack', 4, 'nsubj:pass')  
('Obama', 1, 'flat')  
('was', 4, 'aux:pass')  
('born', 0, 'root')  
('in', 6, 'case')  
('Hawaii', 4, 'obl')  
('.', 4, 'punct')  
('He', 3, 'nsubj:pass')  
('was', 3, 'aux:pass')  
('elected', 0, 'root')  
('president', 3, 'xcomp')  
('in', 6, 'case')  
('2008', 3, 'obl')  
('.', 3, 'punct')
```

spaCY

- See Notebook [https://github.com/kjmazidi/NLP/blob/master/Part 3-Sentences/Chapter 10 Parsing/Dependency parse spaCy.ipynb](https://github.com/kjmazidi/NLP/blob/master/Part%203-Sentences/Chapter%2010%20Parsing/Dependency%20parse%20spaCy.ipynb)

Practice

- The quick brown fox jumped over the lazy river.



Essential points to note

- NLP practitioners have three types of sentences syntax parsers:
- PSG
- Dependency
- SRL
- And many NLP toolkits to use
- Sentence parsing is an important preprocessing step for many NLP applications

Next topic

Documents

