

# 6.1. Dependency Parsing

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# Words and Entities

- Previously, we used sequence labelling to label named entity spans
- Recognising entities means processing the **meaning** of a piece of text, e.g.:

“United Airlines said Friday it has increased fares by \$6.”

- Which view of meaning do we use for NER?

Relational: the relationships between words encode meaning

Compositional: meaning is formed by combining units of text

Contextual: understanding a word from its context

# Words and Entities

- Previously, we used sequence labelling to label named entity spans
- Recognising entities means processing the **meaning** of a piece of text, e.g.:

“United Airlines said Friday it has increased fares by \$6.”
- HMM and CRF rely on the context of a word – its neighbours in a sentence

Relational: the relationships between words encode meaning

Compositional: meaning is formed by combining units of text

Contextual: understanding a word from its context

# Words → Syntax

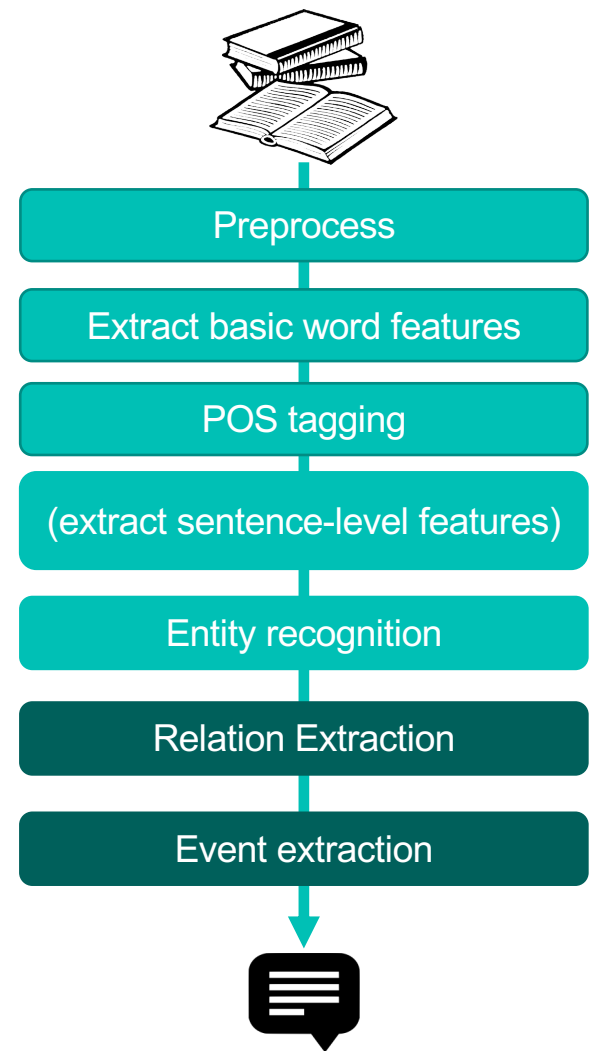
“United Airlines said Friday it has increased fares by \$6.”

- There is a lot more information in here that we can only extract by analysing the **composition** of the whole sentence.
- The arrangement of words to form phrases and sentences is known as **syntax**
- We need an understanding of syntax to understand the full **meaning** of the sentence.

# Words → Syntax → Meaning

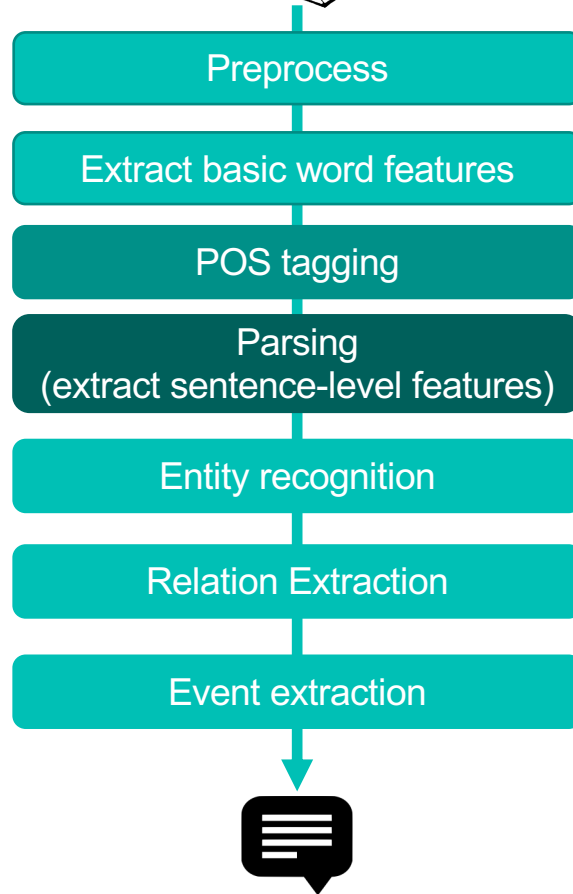
“United Airlines said Friday it has increased fares by \$6.”

- Who increased fares?
- How much did they increase by?
- To answer this, we need to recognise how “United Airlines” and “\$6” relate to the verb “increased”



# Parsing

- Analysing the syntax of a sentence is known as **parsing**
- Parsing requires part of speech tags
- Parts of speech are categories of words that can be used in a particular way to construct phrases and sentences

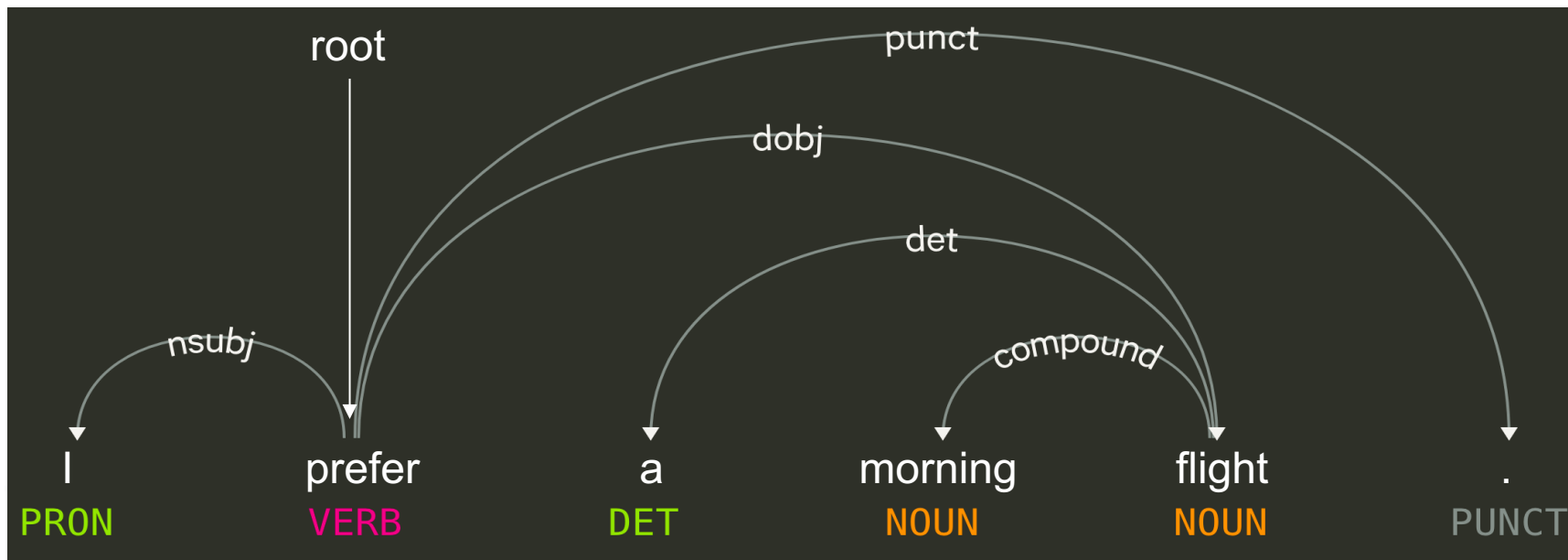


# Chunking

- Chunking splits a sentence into phrases, e.g.:
  - [NP *The morning flight*] [PP *from*] [NP *Denver*] [VP *has arrived.*]
  - Chunk tags are useful features for NER
  - <https://www.nltk.org/api/nltk.chunk.html>
- But chunks don't represent the whole structure of the sentence, so don't give us all we need for tasks like relation extraction.
- Chunking is not suitable for many languages:
  - When phrases can be embedded in one another (e.g., German)
  - If word order is very variable (e.g., Czech, Persian)

# Dependency Parsing

- Extract a graph representing the sentence structure
- <https://explosion.ai/demos/displacy>

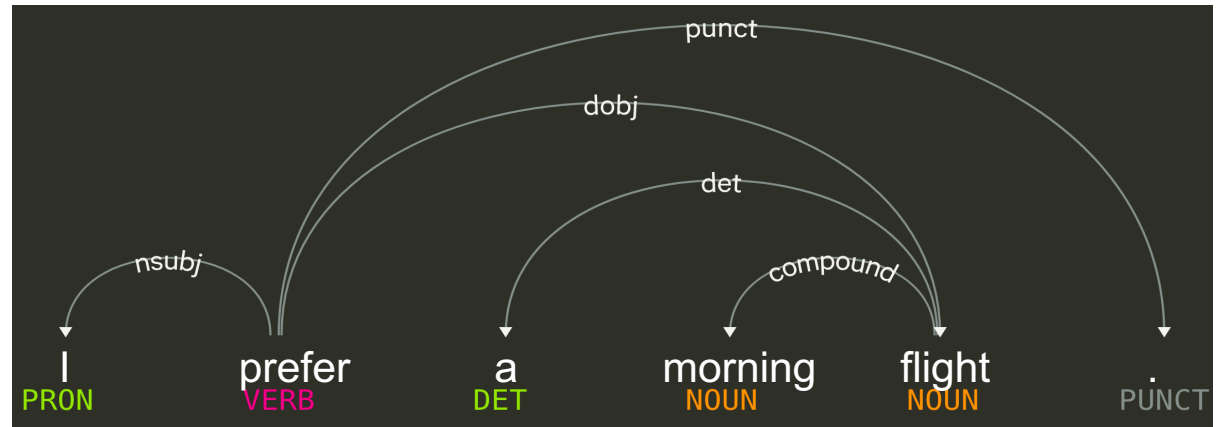




# Dependency Parsing

- *prefer* → nsubj → *I*
- *prefer* → dobj → *flight*
- *flight* → det → *a*
- *flight* → compound → *morning*
- *prefer* → punct → *.*

Each word has one incoming edge and one ancestor, called the *head*



# Relations: Verbs

Relation type	Description	Sentence	Relation triple
NSUBJ	Nominal subject	I <i>prefer</i> a morning flight.	<i>prefer</i> → nsubj → I
DOBJ	Direct object	I <i>prefer</i> a morning <b>flight</b> .	<i>prefer</i> → dobj → <i>flight</i>
IOBJ	Indirect object	We <i>booked</i> <b>her</b> a flight.	<i>booked</i> → iobj → <i>her</i>

Subset of relations from: Nivre, J., et al. (2016).  
Universal Dependencies v1: A multilingual treebank  
collection. LREC

# Relations: Modifiers

Relation type	Description	Sentence	Relation triple
NMOD	Nominal modifier	I prefer a <b>morning</b> <i>flight</i> .	<i>flight</i> → nmod → <i>morning</i>
AMOD	Adjectival modifier	Book the <b>quickest</b> <i>train</i> .	<i>train</i> → amod → <i>quickest</i>
NUMMOD	Numeric modifier	They booked <b>4</b> <i>seats</i> .	<i>seats</i> → nummod → 4

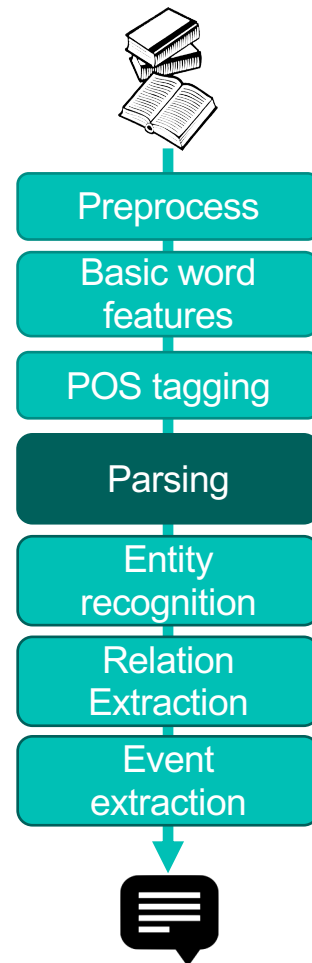
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# Advantages of Dependency Parsing

- Dependency parsing is effective on languages with variable (free) word order
- Dependencies between verbs and their subjects and objects help identify relations between entities in information extraction
- Modifiers change the meaning of words, e.g., “morning flight” makes the word “flight” more specific
- Dependencies therefore encode vital information for answering questions, such as “which flight does the customer prefer?”

# From Parse Trees to Features

- Choice of features is highly task-specific
- Features for a word,  $w$ :
  - Ancestor word (head)
  - Type of the incoming dependency
- To classify relations between two entities, features may include:
  - Head word of each entity (highest word in the parse tree)
  - Dependency types along the path from one head to another
  - Dependency path length
  - Distance to common ancestor



# Challenges of Dependency Parsing

- Goal: Find the dependency relations that make up the sentence
- What are the main challenges of this task?

**Ambiguity: multiple possible parses for a sentence.**

**Fluidity of grammar: sentences that break the rules!**

# Summary

- Understanding syntax helps us process the meaning of sentences
- Dependency parsing produces a graph of relations between words
- Dependency relations provide features for downstream tasks
- For example, relations between verbs and their subjects and objects help us identify relations between entities.