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CSCE 771: Computer Processing of Natural Language

Lecture 6: Shallow Parsing, Dependency Parsing, Project Review

PROF. BIPLAV SRIVASTAVA, AI INSTITUTE

5TH SEPTEMBER, 2024

Carolinian Creed: "I will practice personal and academic integrity."

Acknowledgement: Used materials by Profs. Mausam, Jurafsky & Martin, Stanford NLP

Organization of Lecture 6

- Opening Segment
  - Review of Last Lecture
- Main Lecture
- Concluding Segment
  - About Next Lecture – Lecture 7

Main Section
- Project: complete reviewing of topics
- Shallow Parsing
- Dependency Parsing

Recap of Lecture 5

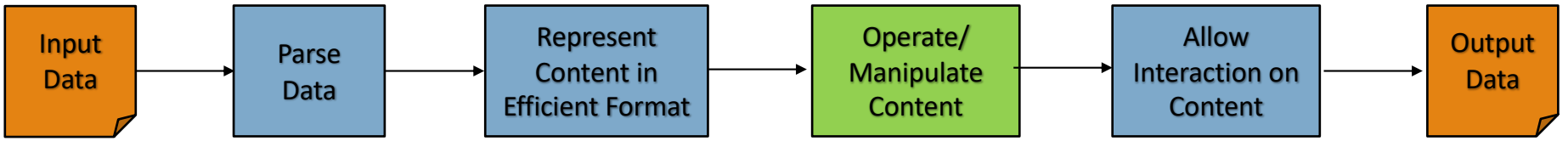
- We discussed the paper - ”Contextual Word Representations: Putting Words into Computers”, by Noah Smith, CACM June 2020”
- We looked at parsing
  - Roles it plays: verifying , generating, recognizing
  - Many types of parsing: shallow parsing for quick NLP tasks, phrase structure parsing, dependency parsing
- Started reviewing projects

Announcements

- Quiz 1 - next class, in-person and using paper and pen
  - No makeup (best of 3 from 4 quizzes)
  - Will cover concepts discussed in class

Main Lecture

Review: Parsing



Parsing

Types of Parsing

- **Phrase structure** / **Constituency Parsing**: find phrases and their recursive structure.
  Constituency - groups of words behaving as single units, or constituents. Context free grammars are also called Phrase-Structure Grammars
  - **Shallow Parsing/ Chunking**: identify the flat, non-overlapping segments of a sentence: noun phrases, verb phrases, adjective phrases, and prepositional phrases.

- **Dependency Parsing**: find relations in sentences

- **Probabilistic Parsing**: given a sentence X, predict the most probable parse tree Y

Chunking

- Chunking - process of identifying and classifying the flat, non-overlapping segments of a sentence that constitute the basic **non-recursive phrases** corresponding to the major content-word parts-of-speech:
  - noun phrases
  - verb phrases
  - adjective phrases, and
  - prepositional phrases

Example

$[_{NP}$ The morning flight] $[_{PP}$ from] $[_{NP}$ Denver] $[_{VP}$ has arrived.]

- Two operations in this type of parsing:
  - segmenting - finding the non-overlapping extents of the chunks and
  - labeling - assigning the correct tag to the discovered chunks
- Some words may not be part of any chunk

From Jurafsky & Martin

Shallow Parsing/ Chunking

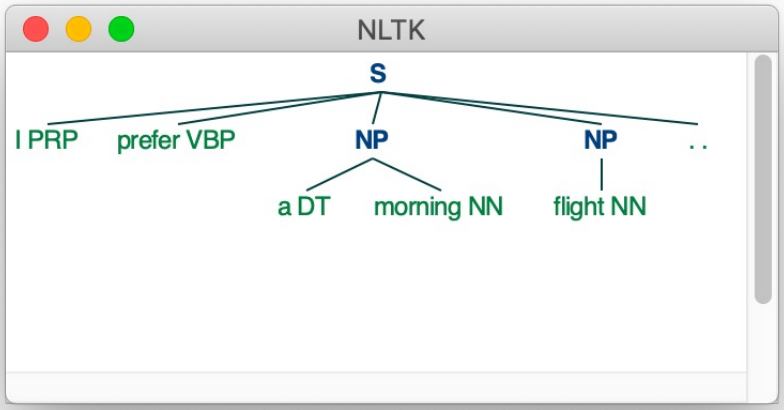
```
data = "I prefer a morning flight."

# Prepare data
tokens = nltk.word_tokenize(data)
tag = nltk.pos_tag(tokens)

# Grammar to use
grammar = "NP: {<DT>?<JJ>*<NN>}"
cp  =nltk.RegexpParser(grammar)

# Parse based on regex
result = cp.parse(tag)
print(result)
```

```
(S I/PRP prefer/VBP (NP a/DT morning/NN) (NP flight/NN) ./.)
```



IOB notation

- Chunking - IOB tagging
  - B - beginning of each chunk type
  - I - inside of each chunk type
  - O - one for tokens outside (O) any chunk
- Total: (2N + 1) tags for N chunk types

Example

The morning flight from Denver has arrived.
B_NP I_NP I_NP O B_NP O O

From Jurafsky & Martin

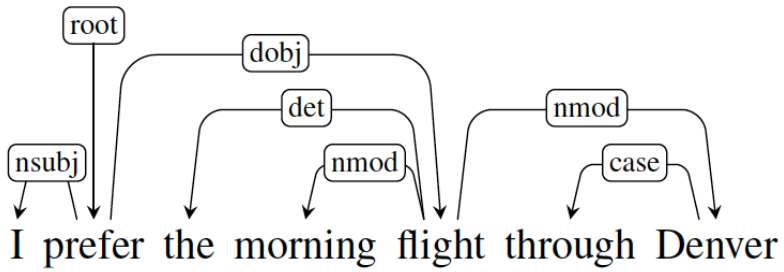
Code and Examples

- Sample code – https://github.com/biplav-s/course-nl-f22/blob/main/sample-code/l6-l7-parsing/Chunking%20-%20syntax%20exploration.ipynb

- Advanced examples – https://www.nltk.org/book/ch07.html

Dependency Parsing

- **Meaning** depends on
  - Words (lemmas) in a sentence
  - Their directed binary grammatical relations with other words
  - (and not on CFGs)
- **Notation**: Labeled arcs are from heads to dependents



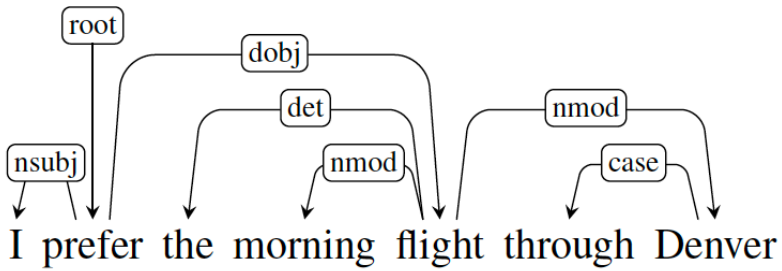
No node corresponding to phrasal constituents or lexical categories in the dependency parse

Dependency Conditions

1. There is a single designated root node that has no incoming arcs.
2. With the exception of the root node, each vertex has exactly one incoming arc.
3. There is a unique path from the root node to each vertex in V.

From Jurafsky & Martin

Dependency Parsing

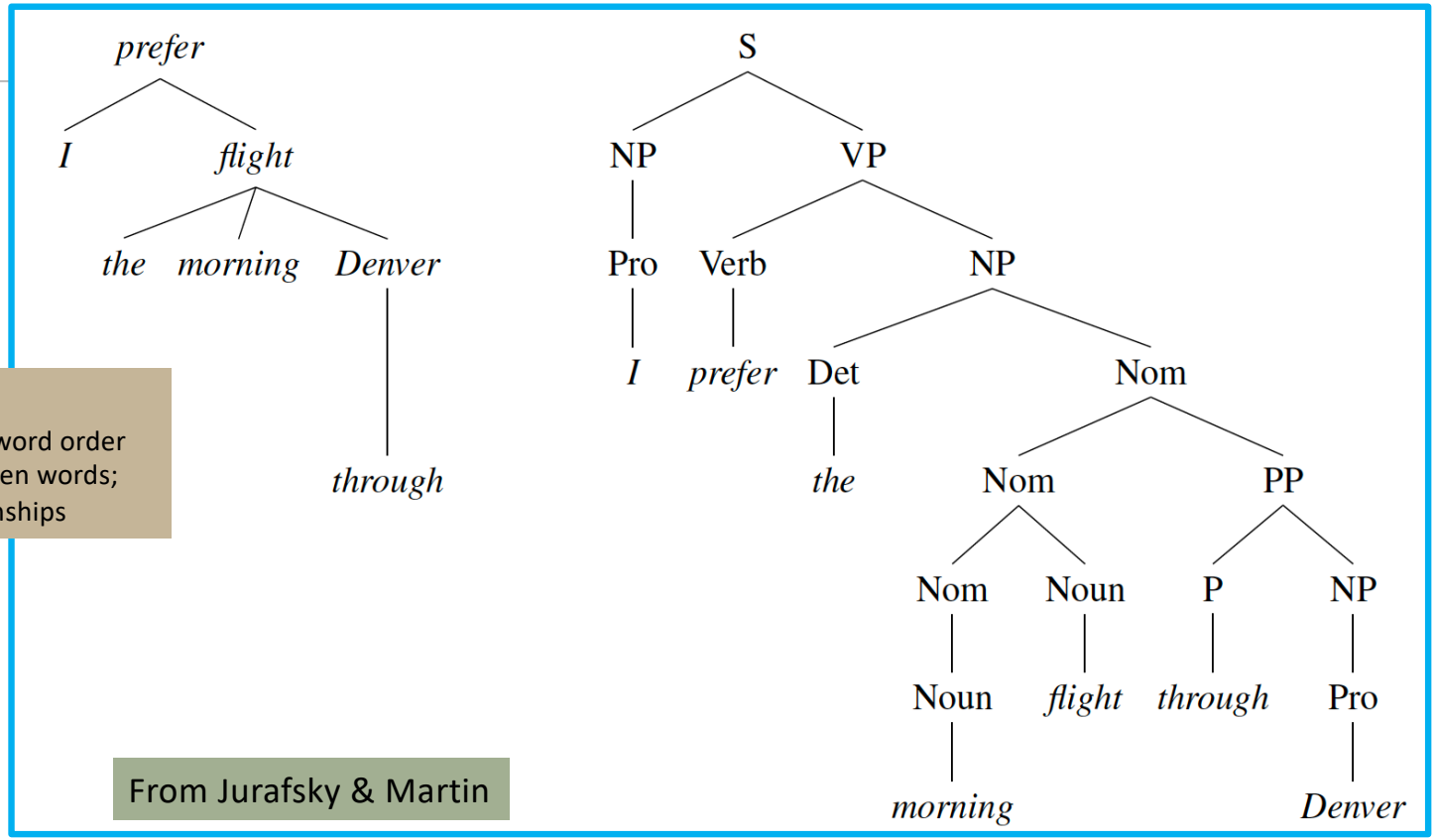


Clausal Argument Relations	Description
NSUBJ	Nominal subject
DOBJ	Direct object
IOBJ	Indirect object
CCOMP	Clausal complement
XCOMP	Open clausal complement
Nominal Modifier Relations	**Description**
NMOD	Nominal modifier
AMOD	Adjectival modifier
NUMMOD	Numeric modifier
APPOS	Appositional modifier
DET	Determiner
CASE	Prepositions, postpositions and other case markers
Other Notable Relations	**Description**
CONJ	Conjunct
CC	Coordinating conjunction

Edge: role that the dependent plays with respect to its head. Examples: subject, direct object and indirect object.

From Jurafsky & Martin

Comparison: Dependency and Phrase Structure

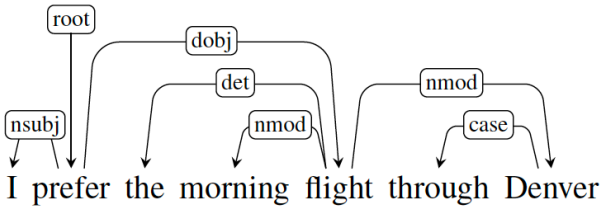


Dependency:

- Useful in languages with free word order
- Highlights relationships between words; useful is tasks needing relationships

From Jurafsky & Martin

Example Dependency Relationships



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From Jurafsky & Martin

Examples of Parsing with Spacy

- Sample code – https://github.com/biplav-s/course-nl-f22/blob/main/sample-code/l6-l7-parsing/parsing%20spacy.ipynb

See GitHub

About Grammar Forms

- **Strong equivalent grammars**: Two grammars are strongly equivalent if they generate the same set of strings and if they assign the same phrase structure to each sentence
- **Weakly equivalent grammars**: Two grammars are weakly equivalent if they generate the same set of strings but do not assign the same phrase structure to each sentence.
- **Chomsky Normal Form**: a grammar which is
  - ϵ-free and
  - each production is either of the form A -> B C or A -> a.

- ***Any context-free grammar can be converted into a weakly equivalent Chomsky normal form grammar***
  - A -> B C D becomes
  - A -> B X and X -> C D

From Jurafsky & Martin

Review: Parsing - CFG

N a set of **non-terminal symbols** (or **variables**)

Σ a set of **terminal symbols** (disjoint from N)

R a set of **rules** or productions, each of the form $A \rightarrow \beta$,
where A is a non-terminal,
β is a string of symbols from the infinite set of strings $(\Sigma \cup N)*$

S a designated **start symbol** and a member of N

Example CFG:

- N = {S, NP, VP, }
- Σ = {he, she, walks, sleeps}
- R = {
  - S -> NP, VP
  - NP -> he
  - NP -> she
  - VP -> walks
  - VP -> sleeps
  }
- S = S

Questions: *which strings are in the language of example CFG*
(a) she sleeps (b) walks sheeps (c) sleeps he (d) she walks (e) he and she walks

Parsing Perspective

- **Question**: Is parsing of a sentence unique ?

Example 1: "*Book the dinner flight*"
- Book the flight which has dinner
- Book the flight for dinner

Example 2: *"I made her duck"*
- I cooked duck (**sense**: **animal**) for her
- I cooked duck (**sense: animal**) belonging to her.
- I turned her into duck (**sense**: **animal**)
- I created duck (**sense: object**) for her
- I made her to lower her head or body (**sense: posture**).

From Jurafsky & Martin

Parsing Perspective

- **Question**: Is parsing of a sentence unique ?
- **Answer**: Not necessarily

Issue: Then, which one to return?

Solution: Given a sentence X, predict its parse tree Y

Lecture 6: Concluding Comments

- We reviewed projects
- We reviewed parsers
  - Shallow parsers
  - Dependency parsers

Concluding Segment

Discussion: Course Project

Theme: Analyze quality of official information available for elections in 2024 [in a state]

- Take information available from
  - Official site: State Election Commissions
  - Respected non-profits: League of Women Voters
- Analyze information
  - State-level: Analyze quality of questions, answers, answers-to-questions
  - Comparatively: above along all states (being done by students)
- Benchmark and report
  - Compare analysis with LLM
  - Prepare report

- Process and analyze using NLP
  - Extract entities
  - Assess quality – metrics
    - Content – *Englishness*
    - Content – *Domain* -- election
  - … other NLP tasks
  - Analyze and communicate overall

Major dates for project check

- Sep 10: written – project outline
- Oct 8: in class
- Oct 31: in class // LLM
- Dec 5: in class // Comparative

Review current states chosen by others

Project Discussion

1. Go to Google spreadsheet against your name
2. Enter the **state** you will focus on for course project

1. Create a private Github repository called "CSCE771-Fall2024-<studentname>-Repo". Share with Instructor (biplav-s) and TA (vr25)
2. Create Google folder called "CSCE771-Fall2024-<studentname>-SharedInfo". Share with Instructor (prof.biplav@gmail.com) and TA (rawtevipula25@gmail.com)
3. Create a Google doc in your Google repo called "Project Plan" and have the following by Friday (Aug 30, 2024)

Timeline

1. Title: Analyze quality of official information available for elections in 2024 in <state>
2. Data need:
   1. Official: state's election commission
   2. LWV: https://www.vote411.org/
3. Methods:
4. Evaluation:
5. Milestones
   - Sep 10: written and feedback
   - Oct 8: in class
   - Oct 31: in class
   - Dec 5: in class

Discussion: Course Project

- Expectations
  - Apply methods learned in class or of interest to a problem of interest
  - Be goal oriented: aim to finish, be proactive, be innovative
  - Do top-class work: code, writeup, presentation
- Typical pitfalls
  - Not detailing out the project, assuming data
  - Not spending enough time
- What will be awarded
  - Results and efforts (balance)
  - Challenge level of problem

Review current states chosen by others

Course Project – Deadlines and Penalty Rubric

- Penalty
  - Missing milestones: **[-10%]**
  - Maximum**: [-40%]**
- Bonus possible
  - if two or more states considered
  -

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About Next Lecture – Lecture 7

Lecture 7

- Statistical parsing
- QUIZ

4	Aug 29 (Th)	NLP Tasks, Case Study – Business Application		
5	Sep 3 (Tu)	Parsing, Paper 1 discussion; project topics review		Practice exercise
6	Sep 5 (Th)	Project topics review, statistic Parsing		
7	Sep 10 (Tu)	Statistical parsing, QUIZ		Quiz 1, **Project Check**
8	Sep 12 (Th)	Evaluation, Semantics		Coding running example
9	Sep 17 (Tu)	Semantics Machine Learning for NLP, Evaluation - Metrics		Code: scikit f1 score package, Code: ConceptIO
10	Sep 19 (Th)	Towards Language Model: Vector embeddings, Embeddings, CNN/ RNN		Code: embedding, genism word vector, tf-idf