

Unit 2

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Syntactic Analysis: Syntactic Representations of Natural Language

- Syntactic analysis, or parsing, focuses on determining the grammatical structure of sentences in natural language.
- It identifies **how words combine to form phrases, clauses**, and complete sentences, adhering to the syntactic rules of a given language.
- Key syntactic representations include:

Constituency-Based Representations

- **Parse Trees:** Hierarchical structures where **nodes** represent **syntactic categories** (e.g., NP, VP), and **leaves** correspond to **words**.
- **Phrase Structure Grammar:** Frameworks like Context-Free Grammars (CFG) underlie these representations.

Dependency-Based Representations

- **Dependency Trees:** Represent relationships between words using **directed edges**, emphasizing **head-dependent structures**.
- **Dependency Parsing:** Efficient for languages with flexible word order.

Hybrid Representations

- Combine **constituency and dependency representations** for richer syntactic insights.

Parse Trees

- A **parse tree** (or **syntax tree**) is a tree structure that **represents the syntactic structure of a sentence** according to a given grammar, typically a **context-free grammar (CFG)**.
- It visualizes **how words** in a sentence are grouped into phrases and how these **phrases relate to one another hierarchically**.
- Parse trees are fundamental tools in computational linguistics and NLP, enabling deeper linguistic insights and effective language processing.

Components of a Parse Tree

- 1.Root Node:** Represents the start symbol of the grammar, often denoted as S (Sentence).
- 2.Non-Terminal Nodes:** Represent intermediate syntactic categories like noun phrases (NP), verb phrases (VP), or prepositional phrases (PP).
- 3.Terminal Nodes:** Represent the actual words (or tokens) in the sentence.

Example

Consider the sentence:

"The cat sleeps."

Using a simplified CFG:

1. $S \rightarrow NP VP$
2. $NP \rightarrow Det N$
3. $VP \rightarrow V$
4. $Det \rightarrow "The"$
5. $N \rightarrow "cat"$
6. $V \rightarrow "sleeps"$

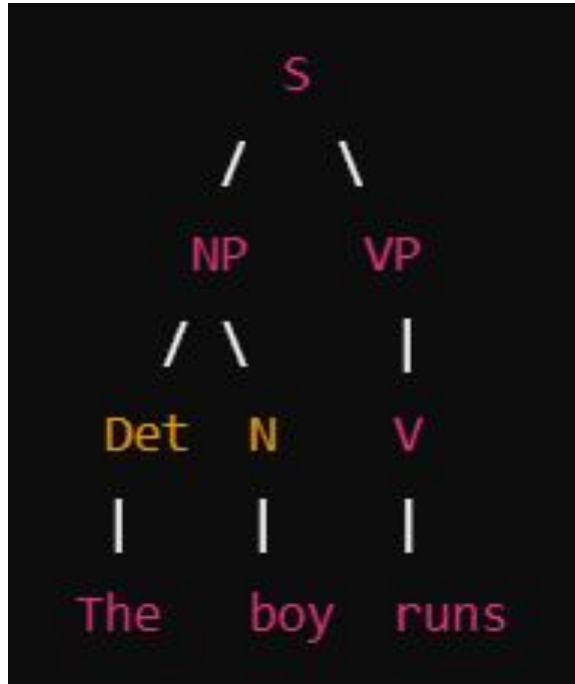
The parse tree looks like this:



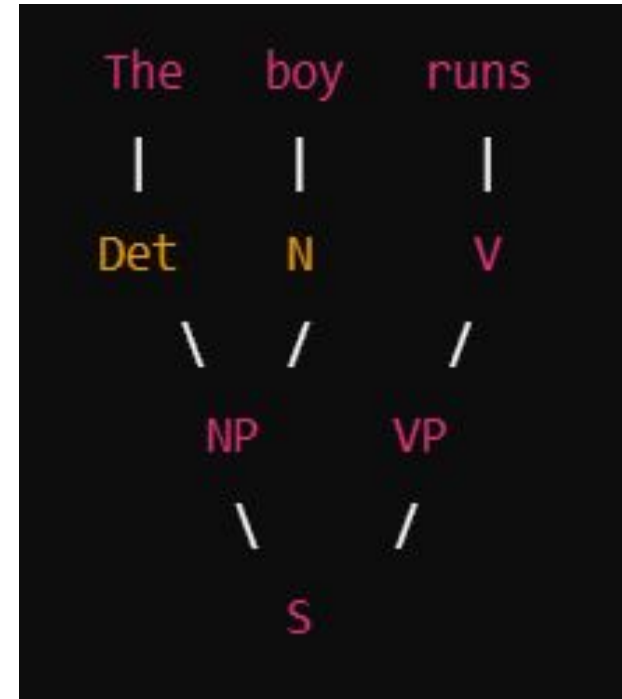
How Parse Trees Are Generated

- Top-Down Parsing:** Start from the root (S) and expand nodes based on grammar rules until terminals match the input sentence.
- Bottom-Up Parsing:** Start with the words (terminals) and combine them using grammar rules until the root (S) is derived.

Sentence: "The boy runs."



- Start with S → Expand into NP (Noun Phrase) and VP (Verb Phrase).
- Expand NP into Det (Determiner) and N (Noun).
- VP expands into V (Verb).



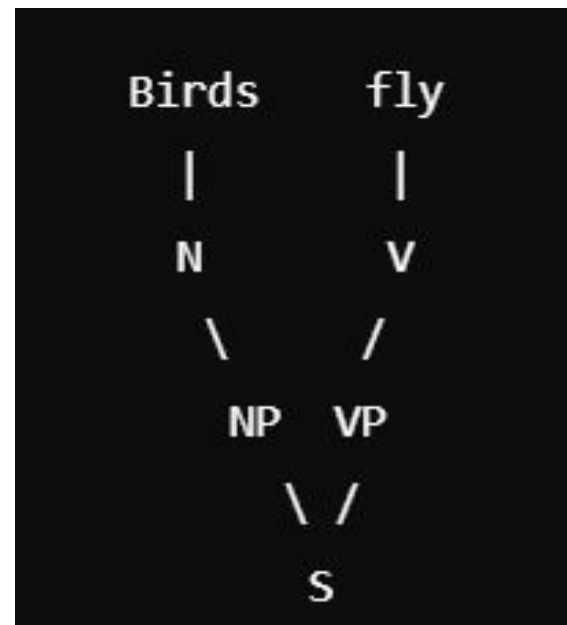
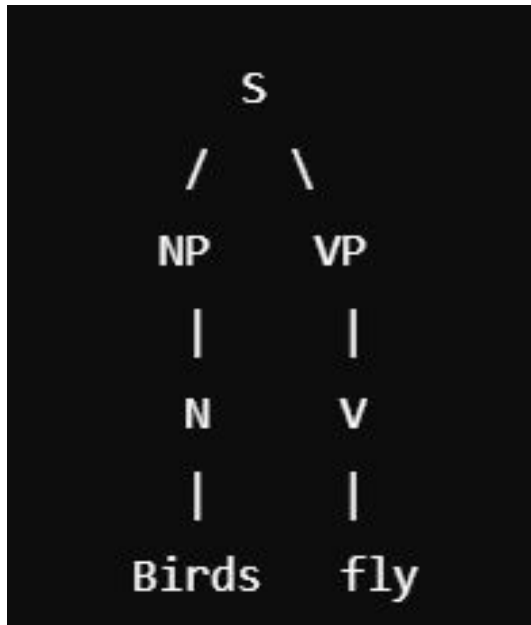
Begin with Det, N, and V → Combine Det and N into NP.

Combine NP and V into S.

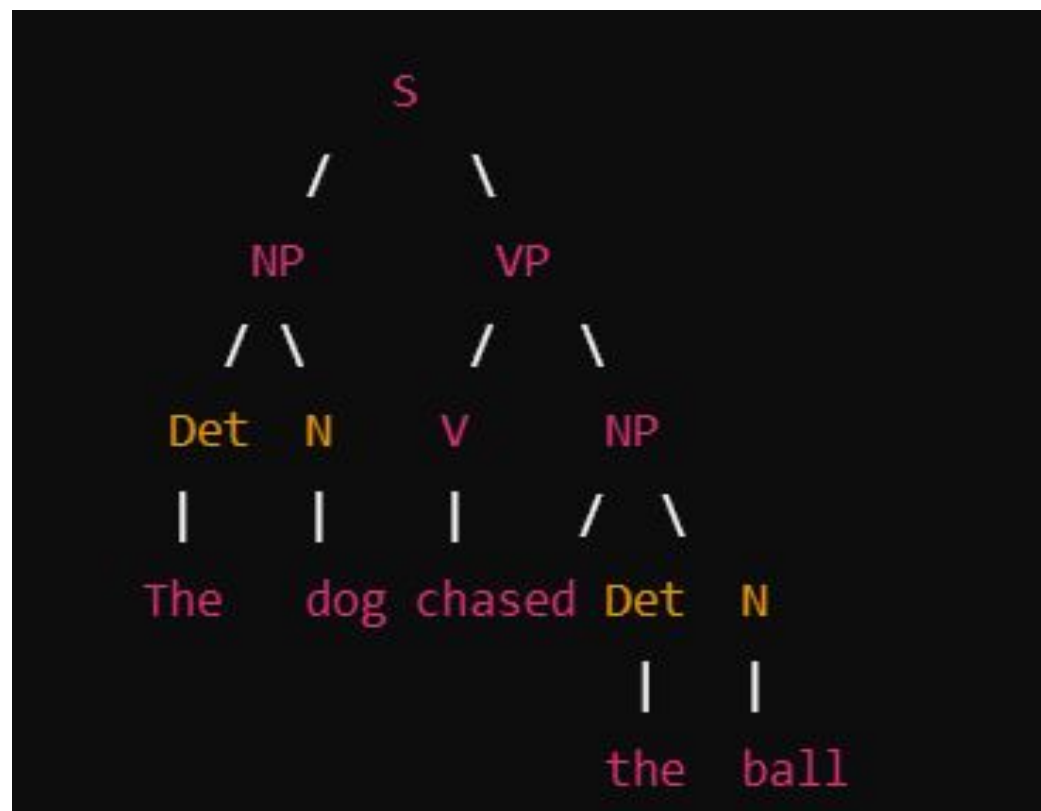
Sentence: "She plays piano."



Sentence: "Birds fly."



Sentence: "The dog chased the ball."



Phrase Structure Grammar

- Phrase Structure Grammar is a **type of formal grammar** used to represent the syntactic structure of sentences. Frameworks like **Context-Free Grammars (CFG)** play a **foundational role** in this representation.
- A set of rules that describe how words and phrases combine to form sentences in a language.
- It defines the hierarchical structure of a sentence by breaking it into smaller constituents or phrases (e.g., noun phrases, verb phrases).

Role of Context-Free Grammars (CFG) in NLP

1. Definition of CFG:

CFG is a type of grammar where **production rules** are used to generate **sentences**.

Each rule has a single non-terminal symbol (like S, NP, VP) on the left-hand side and a combination of terminal and non-terminal symbols on the right-hand side.

$$S \rightarrow NP VP$$
$$NP \rightarrow Det N$$
$$VP \rightarrow V NP$$

2. Tree Representation:

- CFG helps **in constructing parse trees**, which visually depict the syntactic structure of a sentence.

Applications in NLP:

- **Parsing:** CFGs are widely used in parsers to analyze sentence structure. Tools like **Stanford Parser** and **spaCy** implement such approaches.
- **Syntax Checking:** CFG ensures that a sentence conforms to the grammatical rules of a language.
- **Machine Translation:** By understanding sentence structure, CFG aids in **generating syntactically correct translations.**
- **Text Generation:** CFG helps in **generating text that adheres to grammatical rules.**

Limitations of CFG in NLP:

- CFGs assume that the structure of a language can be fully captured by context-free rules, but **natural languages often exhibit dependencies** (e.g., subject-verb agreement, long-distance dependencies) that are **not easily handled by CFG alone**.
- Extensions like **Lexicalized CFGs** or **Tree-Adjoining Grammars (TAGs)** are used to address these complexities.

Dependency-Based Representations

- Dependency-based representations **focus on the syntactic relationships** between words in a sentence. Instead of relying on **phrase structure (like in constituency-based parsing)**, dependency representations **emphasize the grammatical dependencies** between individual words.

- **Key Features:**

1. Head-Dependent Structure:

1. Each word in a sentence (except the root) is dependent on another word, which acts as its **head**.
2. For example, in the sentence "*She eats an apple*", the word "*eats*" is the **head**, and "*She*", "*an*", and "*apple*" are its dependents.

Directed Graph Representation:

Sentences are represented as directed graphs or trees, where:

- Nodes correspond to words.
- Edges represent dependencies (e.g., subject, object, modifier).

Grammatical Relations:

Dependencies are labeled with **grammatical relations** such as subject (*nsubj*), object (*obj*), and modifier (*amod*).

Non-Projectivity:

Some languages (e.g., German) exhibit non-projective dependencies, where dependencies can cross, making them more challenging to parse.

Dependency Trees

- A dependency tree is a tree structure that represents the syntactic structure of a sentence **based on dependency relations**.

Characteristics:

1. Single Root:

- The tree has a single root node, **typically the main verb** or **predicate of the sentence**.
- Example: In *"She quickly eats an apple"*, *"eats"* is the root.

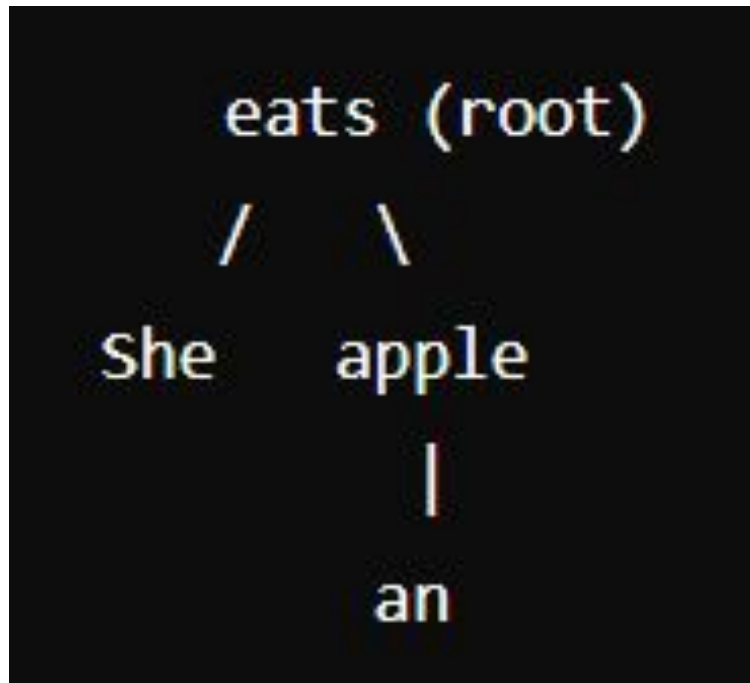
2. Child Nodes:

- Words **directly dependent on the root** or other nodes are represented as child nodes.
- Each word (except the root) has **exactly one parent**.

3. Labeled Arcs:

- Arcs (edges) between words **are labeled with dependency types** (e.g., subject, object).

For the sentence "*She eats an apple*", the dependency tree might look like this:



- *She* → *eats*: Subject relation (*nsubj*).
- *eats* → *apple*: Object relation (*obj*).
- *apple* → *an*: Determiner relation (*det*).

Dependency Parsing

- Dependency parsing is the process of analyzing the grammatical structure of a sentence by **identifying dependency relationships between words** and constructing a dependency tree.

- **Types of Dependency Parsing:**

1. Projective Parsing:

- Assumes **no crossing dependencies**.
- Works well for languages with **simple structures** (e.g., English).

2. Non-Projective Parsing:

- Allows crossing dependencies.
- Necessary for languages like **Czech or German** with complex word order.

Parsing Methods:

1. Rule-Based Parsers:

- Use hand-crafted rules to determine dependencies.
- Example: Grammar-based parsers.

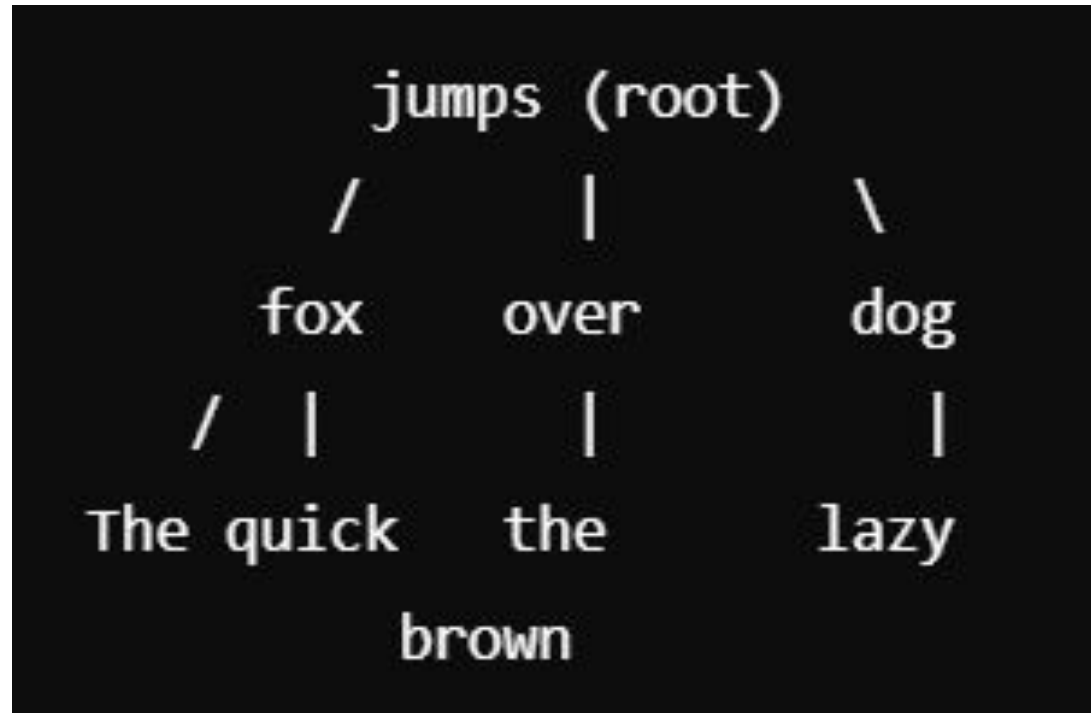
2. Statistical Parsers:

- Use machine learning to predict dependencies based on training data.
- Example: Transition-based parsers and graph-based parsers.

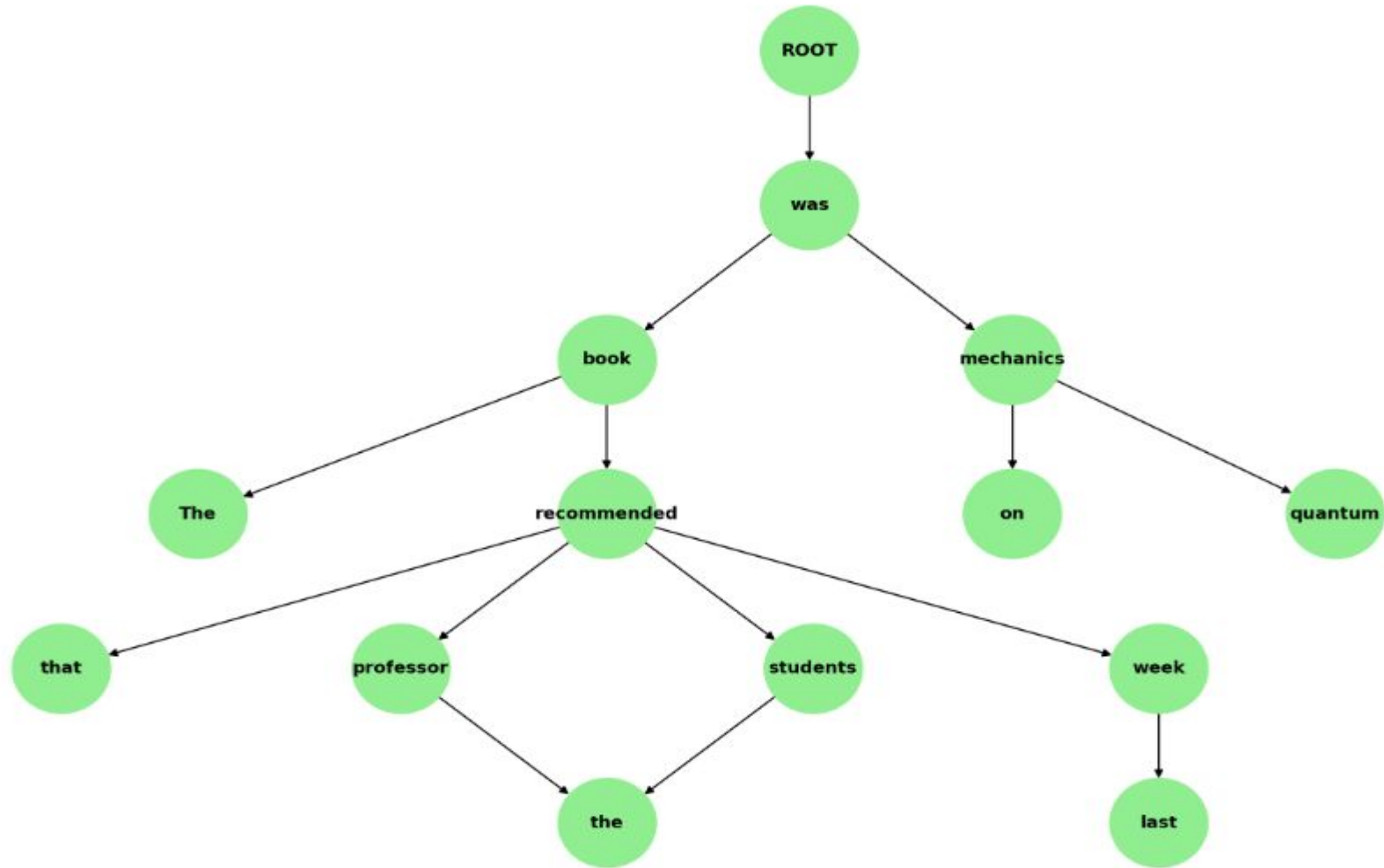
3. Neural Parsers:

- Leverage deep learning models for dependency parsing.
- Example: BERT-based parsers.

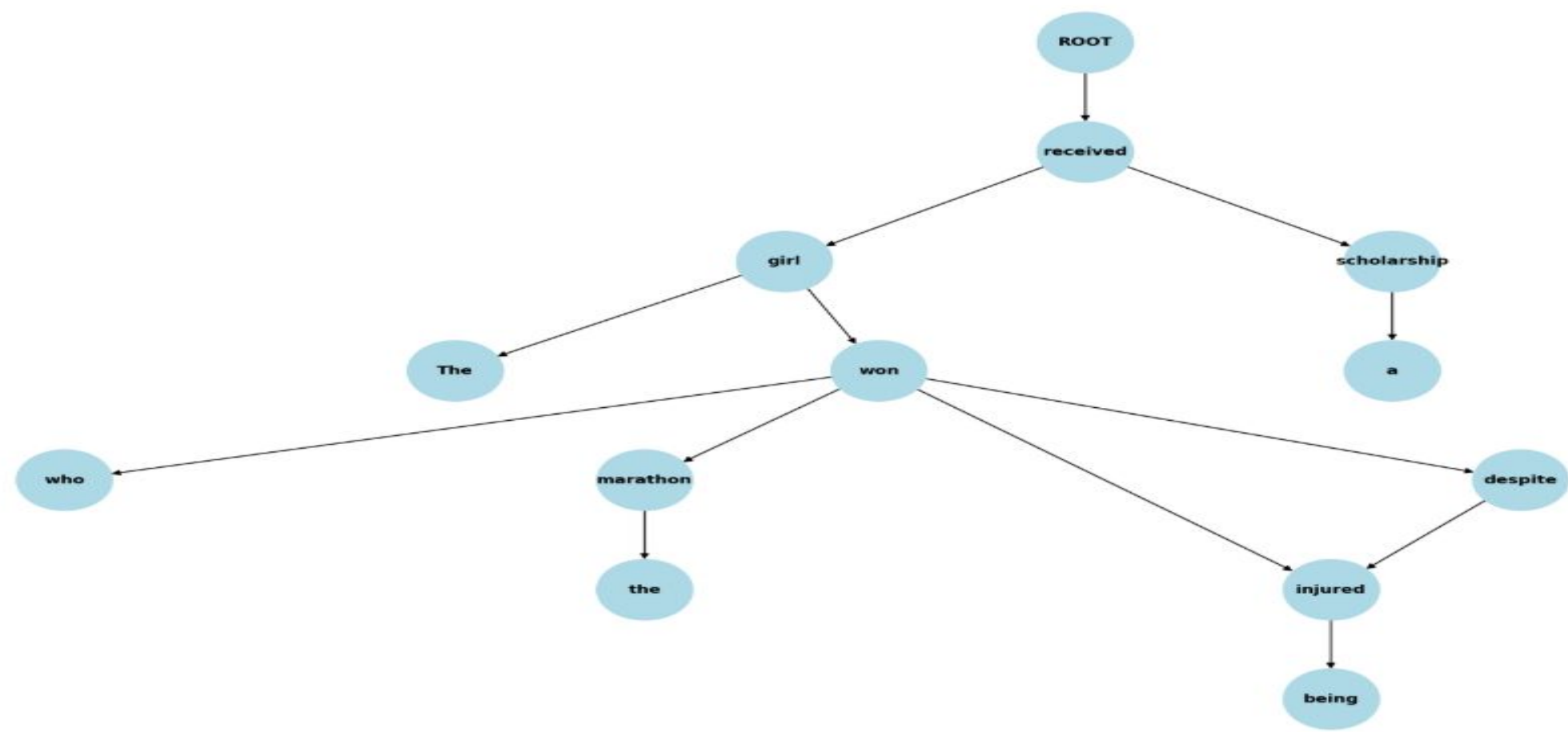
For the sentence: "*The quick brown fox jumps over the lazy dog*", a dependency tree would look like:



"The book that the professor recommended to the students last week was on quantum mechanics."



"The girl who won the marathon despite being injured received a scholarship."



- Dependency tree generation involves algorithms that analyze the syntactic structure of sentences based on the relationships between words. These algorithms fall under the domain of **dependency parsing** in Natural Language Processing (NLP).

Transition-Based Parsing

- Transition-based parsers incrementally build dependency trees by performing a sequence of actions (e.g., SHIFT, REDUCE).
- **Key Algorithms:**
 - **Arc-Standard Parser:** Builds dependency arcs bottom-up.
 - **Arc-Eager Parser:** Builds arcs as soon as possible (top-down approach).
 - **Arc-Hybrid Parser:** Combines aspects of both Arc-Standard and Arc-Eager.
- **Popular Libraries:** SpaCy, Stanford CoreNLP.

- **Common Libraries and Tools**

1. **SpaCy**: Lightweight, fast, and uses transition-based parsing.
2. **Stanza (StanfordNLP)**: Neural pipeline supporting dependency parsing.
3. **UDPipe**: Pre-trained models for Universal Dependencies datasets.
4. **MaltParser**: Implements transition-based dependency parsing.
5. **ZPar**: A statistical dependency parsing library.

Projective vs. Non-Projective Dependency Trees

- Dependency trees can be classified as **projective** or **non-projective** based on whether the edges cross when drawn over the linear sequence of words in a sentence.

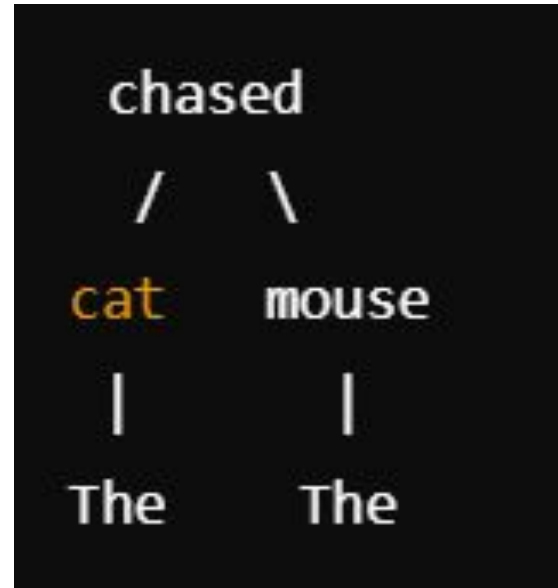
1. Projective Dependency Tree

In a projective tree:

- All edges can be drawn without crossing each other.
- This structure is typical for strict word-order languages like English.
- **Example Sentence (Projective):**
"The cat chased the mouse."

Dependency Tree:

- chased → cat (subject)
 - chased → mouse (object)
 - cat → The (determiner)
 - mouse → The (determiner)
-
- The cat chased the mouse
 - The mouse that the cat chased ran away. (Non Projective)



2. Non-Projective Dependency Tree

In a non-projective tree:

- At least one edge crosses another.
- This structure occurs in free-word-order languages like Hindi, Russian, or Czech.

Example Sentence (Non-Projective):

"The report, I told you yesterday, is on the table."

Dependency Tree:

- **is** → **report** (subject)
- is → on (prepositional phrase)
- on → table (object of preposition)
- **report** → **told** (relative clause)
- told → I (subject of told)
- told → yesterday (temporal modifier)
- told → you (object)

