

Module-IV

Semantic & Discourse level Analysis

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Semantic Analysis

Semantic Analysis

- Semantic analysis involves mapping of natural language utterances to some representation of meaning.
- Semantic analysis creates a representation of the meaning of a sentence,
 - where the basic building blocks of the system are: Entities, Concepts, Relations, Predicates
- Many language processing tasks requires non linguistic knowledge of the words besides, morphological and syntactic knowledge.
- A semantic representation bridges the gap between the linguist and common sense knowledge.
- The language used to describe the semantics representations is called a **meaning representation language**

Meaning Representation

- A meaning representation language bridges the gap between linguistic and common sense knowledge.

Example:

A cup on the table

On(cup, table)

Air India serves Hyderabad

Serves(Air India, Hyderabad)

Cont...

- Requires semantic knowledge
 - Word meanings
 - How word meanings combined in sentences to form sentence meanings
- Examples of meaning representation languages:
 - **First order predicate logic (FOPL)**
 - Semantic Nets
 - Conceptual dependency

Characteristics of a Meaning Representation Language

- A meaning representation language must be
 - **Verifiable**
 - **Unambiguous**
 - **Supports canonical forms**
 - **Support inference and variables**
 - **Expressiveness**

Cont...

Verifiability:

- Determines the truth of the representation in the knowledge base
- Example:
 - Does Kingfisher serve Hyderabad?
 - Serves(Kingfisher, Hyderabad)

Unambiguous:

- Support representations that have only one possible interpretation

Cont...

Support for canonical form:

Examples:

Does Air India **offers** a flight to Hyderabad

Does Air India **have** a flight to Hyderabad

Does Air India **serves** Hyderabad

Meaning representation:

Serves(Air India, Hyderabad)

Serves: {offers, have, serves}

Cont...

Support inference and variables:

Example:

Serves(Air India, Hyderabad)

$R(x, y)$

Expressiveness:

Should be able to address different domains

Example:

bats in the field

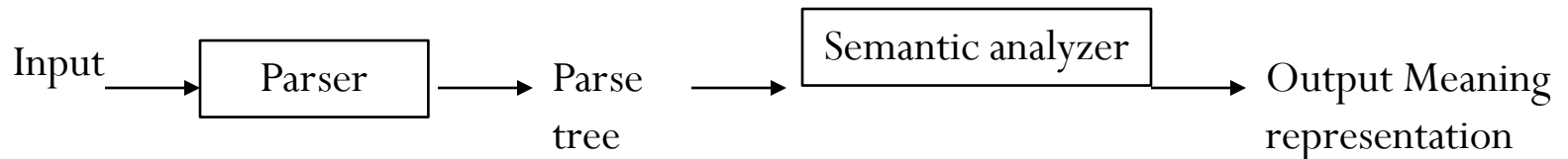
Approaches to Meaning Representation

1. Syntax-driven semantic analysis

2. Semantic grammar

1. Syntax-driven Semantic Analysis

- Meaning of sentence are obtained from meanings of parts
 - E.g. groupings and relations from syntax
- It uses syntactic constituents of a sentence to build its meaning representation



Example:

<u>President</u>	<u>nominates</u>	<u>speaker</u>
N	V	N
[_NP___]	[_____VP_____]	

S -> NP VP

Cont...

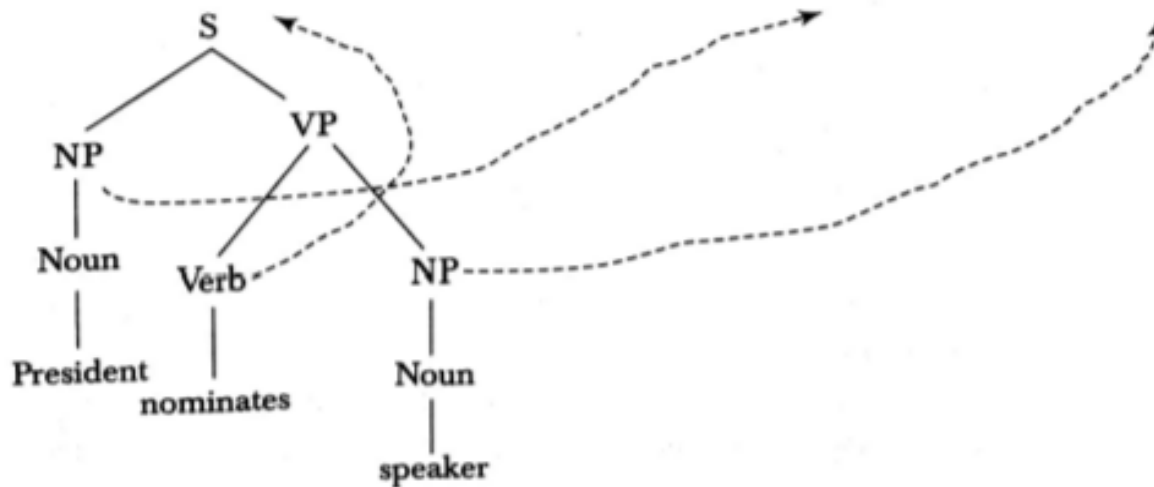
$\exists e, x, y \text{ is_a } (e, \text{nomination}) \wedge \text{nominator } (e, x) \wedge \text{nominee } (e, y)$

$\text{verb} \rightarrow \text{nominates } \{ \exists e, x, y \text{ is_a } (\text{nomination}) \wedge \text{nominator } (e, x) \wedge \text{nominee } (e, y) \}$

$\text{VP} \rightarrow \text{verb NP} \{ \text{verb}_{\text{.sem}} (\text{NP}_{\text{.sem}}) \}$

President nominates speaker.

$\exists e \text{ is_a } (e, \text{nomination}) \wedge \text{nominator } (e, \text{President}) \wedge \text{nominee } (e, \text{speaker})$



2. Semantic Grammar

- Syntax-driven semantics analysis relies on traditional grammar rules which are not meant for semantic processing.
- Semantic grammar brings semantic capabilities into the grammar
 - A grammar developed with the intension of handling semantics
- Rules and constituents corresponds directly to entities and activities in the domain

Example:

I want to go from Delhi to Chennai on 24th April

- A rule can be created of the form:
InfoRequest: User wants to go <to city> <from city> <TimeExpr>

Lexical Semantics

- It involves studying the meaning of individual words
- It includes words, sub-words, affixes (sub-units), compound words and phrases.
- Lexical semantics plays an important role in semantic analysis, allowing machines to understand relationships between lexical items (words, phrasal verbs, etc.)
- The relationships are useful in organizing words in lexical databases.
- **WordNet** is a widely known lexical database

Types of Word Relationships

- **Hypernym:**

- A hypernym is a word with a more general sense.

Example:

- ‘Automobile’ is a hypernym for ‘car’, ‘truck’

- **Hyponym:**

- A hyponym is a word with a more specific meaning.

Example:

- ‘car’ is hyponym of ‘automobile’

- **Polysemy:**

- Word with different but related meanings

Example: ‘chair’

‘chair person’ or ‘furniture’

Cont...

- **Homonyms:**

- Two words that are sound the same and are spelled alike but have a different meaning.

Example: bank (river bank), bank (Financial institution)

- **Synonyms:**

- Words that have the same sense or nearly the same meaning as another
- Examples: happy, ecstatic, overjoyed

- **Antonyms:**

- Words that have opposite meanings

Example: happy, sad

WordNet

- **WordNet** is a large lexical database for English language that lists all senses of a word.

Example:

“read”

Noun:

1. read (something that is read) “The article was a very good read”

Verb:

1. read(interpret something that is written or printed)
“read the newspaper”
2. Read (scan) “this dictionary can be read by the computer”
3. Read (interpret the significance of; human behavior)
“She read the sky and predicted rain”;
“I cant read his strange behavior”

....

11.

Internal Structure of words

Thematic role:

- Thematic role is the semantic relationship between predicate (a verb) and an argument (noun phrase) of a sentence

Example:

Nomination -> nominator

Singing -> singer

Cont...

Selectional Restrictions:

- Selectional restrictions are semantic constraints that are placed on arguments for a given word sense.

Example:

eat requires some eating object

A table eats grass **X**

$\exists e, x, y$ eating(e), \wedge
agent(e, x) \wedge
theme(e, y)

Ambiguity

- Ambiguity refers to a situation where an expression (words or phrases or sentence) can have more than one interpretation.
- Natural languages are highly ambiguous and resolving the varied interpretations in a reasonable time by computerized models is a difficult task.
- Ambiguity can occur at four different levels:
 - Lexical
 - Syntactic
 - Semantic
 - Pragmatic

Lexical Ambiguity

- Lexical ambiguity is the ambiguity of a single word.
- A word may be ambiguous with respect to its internal structure or its syntactic class.

Examples:

book (Noun, Verb)

plant, bat

Syntactic Ambiguity

- Syntactic ambiguity is the ambiguity in grammatical structure leading to different interpretations.

Examples:

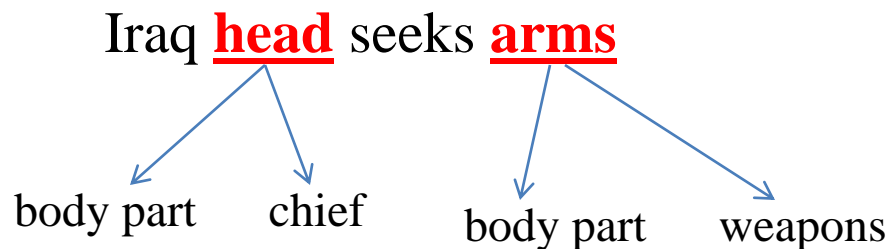
“He saw the man with a telescope”

“Stolen painting found by tree”

Semantic Ambiguity

- Semantic ambiguity occurs when meaning of the words can be misinterpreted.
- i.e. the meaning of the words in a phrase can be combined in different ways, leading to different interpretations.
- Different interpretations of the same sentence.

Example:



Pragmatic Ambiguity

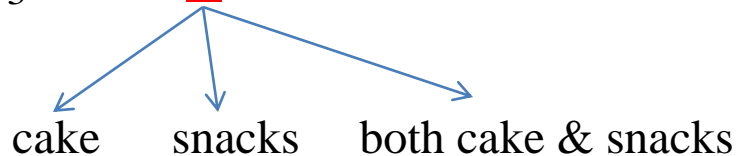
- Pragmatic Ambiguity refers to a situation where the context of a phrase gives it multiple interpretations.

Example:

S₁: Cake is on the table.

S₂: I have prepared some snacks

S₃: Give it to the kids



Cont...

S₁: Cake is on the table.

S₂: I have prepared some snacks

S₃: Give it to the kids

S₄: Kids enjoyed cake and snacks

=> 'it' refers to "both cake and snacks"

Word Sense Disambiguation (WSD)

- WSD is the process of identifying which sense of a word is used in a sentence when the word has multiple meaning.
- Given a word and its possible senses as defined by a dictionary, the objective is to classify an occurrence of the word in context into one or more of its sense classes.

Example:

- The word “*orange*” can refer to a color, a fruit, or even a city in Florida
- WSD has its applications in Machine Translation systems, question answering systems, text to speech systems, etc.

Approaches to WSD

- **Selectional Restriction-based WSD**
- **Context-based WSD**

Selectional Restriction-based WSD

- Selectional restrictions or preferences can be used in parsing to eliminate incorrect meaning representations

Example:

employ

The institute will employ new employees. ('to hire')

The committee employed her proposal. ('to accept')

- Human/organization -> human(to hire)
- Human/organization -> idea(to accept)

Context-based WSD

- Uses the context of ambiguous word
- Two types:
 - **Knowledge-based**
 - **Corpus-based**

Knowledge-based WSD

- It utilizes information from an explicit lexicon or knowledge base to disambiguate a word.
- The lexicon or knowledge base used may contain machine readable dictionary or hand coded rules to disambiguate.

Corpus-based WSD

- A corpus-based approach extracts information on word sense from a large sense tagged corpus.
- The information used to annotate an ambiguous word may be:
 - Distributional information (frequency distribution of a sense)
 - Context
 - Additional knowledge sources
- Two approaches to corpus-base WSD
 - 1. Supervised WSD**
 - 2. Unsupervised WSD**

Cont...

Supervised WSD:

- Rely on a sense-tagged training corpus for disambiguation.
- Classification algorithms like KNN may be applied for disambiguation.
- With a good quality tagged data set, the overall accuracy in WSD can be very high.

Unsupervised WSD:

- Uses raw or unlabelled text corpora for training and annotated data for evaluation.
- A completely unsupervised algorithm can not perform sense tagging. It can only perform sense discrimination.
- The unsupervised WSD algorithm creates a number of clusters based on the context of an ambiguous word and discriminate between them without labeling them.

Knowledge Sources in WSD

- Context of a word
- Frequency of a sense
- Selectional preferences
- Domain

Issues in Processing Indian Languages

- Indic scripts have a nonlinear structure
- Have a free word order
 - i.e. words can be moved freely within a sentence without changing the meaning of the sentence
- Have rich set of morphological variants
- Extensive and productive use of complex predicates
- Ambiguity

Discourse Analysis

Discourse Analysis

- Attempts to interpret structure and meaning of larger units
 - at paragraph and document level, in terms of words, phrases, clusters and sentences.
- Requires **Discourse Knowledge**
 - Knowledge of how the meaning of a sentence is determined by processing sentences
- Concerned with linguistic phenomena and extra linguistic phenomena
- Discourse level analysis uses: the situational context, cultural, social, interpersonal and linguistic knowledge

Cohesion

- Cohesion bounds text together.
- It's the textual phenomenon that involves linking elements

Example:

- “Yesterday, my friend invited me to my friend's house. When I reached, my friend was preparing coffee. My friend's father was cleaning dishes. My friend's mother was busy writing a book”
- This type of over specification is avoided through the use of “her”
- “Yesterday, my friend invited me to her house. When I reached, my friend was preparing coffee. Her father was cleaning dishes. Her mother was busy writing a book”
- “Her” is cohesive with “my friend”

Reference

- Link a referring expression to another referring expression in the surrounding expression.

Example:

- Suha bought a printer. **It** cost **her** Rs.20000.
 - ‘Her’ refers to a person named ‘Suha’
 - ‘It’ refers to an entity named ‘printer’

Ellipsis

- It's a form of grammatical cohesion.
- It refers to the phenomenon when a part of a sentence is omitted or left unpronounced.
- The reader uses the surrounding text to recover the omitted text

Example:

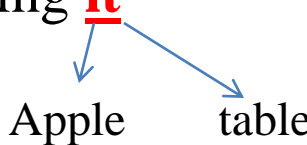
- I know that lady. Do you?
- Here 'Do You' is an ellipsis stands for 'Do you know that lady'

Reference Resolution

- It is the process of resolving references to earlier or later items in the discourse.
- Reference resolution techniques use various constraints and preferences to identify the preferred reference.

Example:

Apple on a table. She is eating it

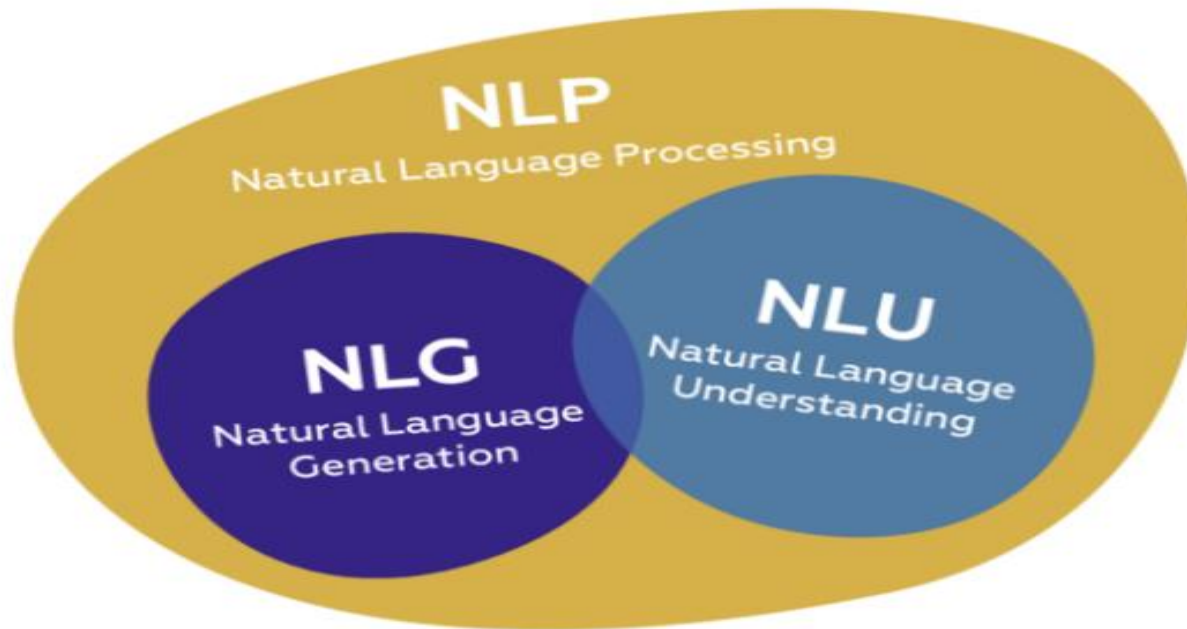


The diagram shows two blue arrows originating from the underlined word 'it'. One arrow points down and to the left towards the word 'Apple', and the other arrow points down and to the right towards the word 'table'.

- Using selectional restrictions 'eat' refers to 'apple' not the 'table'

Natural language Generation (NLG)

The two Components of NLP



- NLU is focused on deriving analytic insights from textual data,
 - NLG is used to synthesize textual content by combining analytic output with contextualized narratives

Natural language Generation (NLG)

- NLG deals with automatic generation of natural language sentence using non linguistic inputs.
- The technology can actually tell a story– exactly like that of a human analyst – by writing the sentences and paragraphs for you.
- NLG is one of the fastest growing technologies being adopted in the industry.
- There many use-cases for NLG, but it is seen to be most effective in automating the time-intensive data analysis and reporting activities.

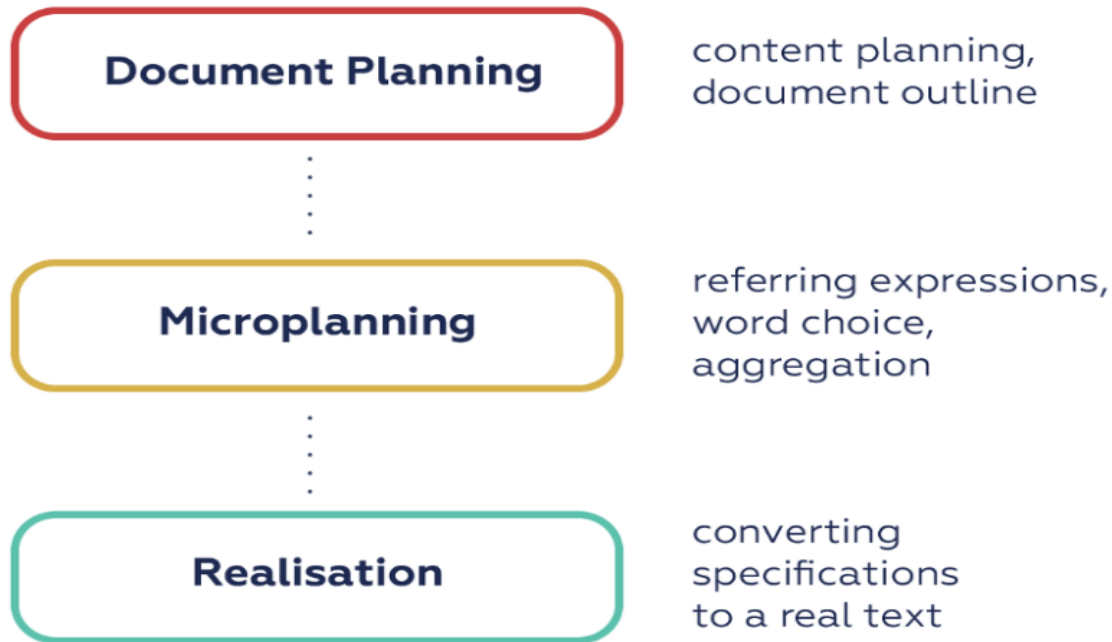
Usable Application of NLG

- Natural language databases: Airline schedule database, Accounting database or spreadsheets, expert system knowledge bases, etc.
- Summary generation
- Language Translator
- Social Media Monitoring
- Chat bots
- Survey Analysis
- Media content creation

NLG Cont...

- The problems with most NLG platforms is that they **hard code intelligence into a template**.
- This makes for systems that are brittle and hard to change and are not able to accept new data without new coding.
- To generate text, the NLG model need to identify:
 - What to write
 - How to write it

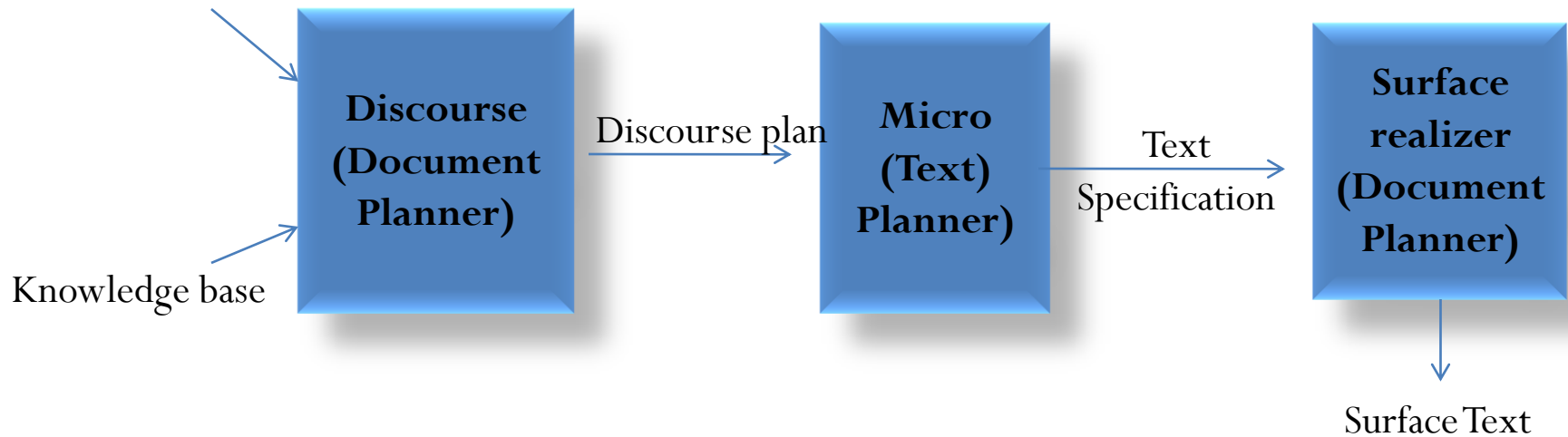
Cont...



- There are two major approaches to language generation: using templates and dynamic creation of documents

Phases of NLG

Communicative goal



Cont...

- **Discourse Planner:**

- Decides the ordering and structure of the text that is to be generated
- Output is represented in the form of a tree

- **Text Planner:**

- Decides words and phrases to express the concepts and structure present in a discourse plan
- Involves sentence aggregation, lexicalization and referencing expression generation

- **Surface Realization:**

- Takes sentence plan produced by the text planner and generate individual sentences

Challenges

- Data availability and quality
- Originality and writing quality
- Bias