

CSCE 771: Computer Processing of Natural Language

Lecture 9: Semantics

PROF. BIPLAV SRIVASTAVA, AI INSTITUTE

15TH SEPTEMBER, 2022

Carolinian Creed: “I will practice personal and academic integrity.”

Acknowledgement: Used materials by
Jurafsky & Martin,

Organization of Lecture 9

- Opening Segment

- Announcement
- Review of Quiz 1
- Project Report Format



- Main Lecture

Main Section

- Semantics
 - Shallow: similarity, relatedness; frames
 - Propbank
 - Deep: AMR
 - ConceptNet
- Review projects

- Concluding Segment

- Reading material:
- About Next Lecture – Lecture 10

7	Sep 8 (Th)	Statistical Parsing, QUIZ
8	Sep 13 (Tu)	Review Parsing, Quiz review, Review Project, Introduce Evaluation
9	Sep 15 (Th)	Semantics
10	Sep 20 (Tu)	Review: Machine Learning for NLP, Evaluation – Metrics
11	Sep 22 (Th)	Language Model – Vector embeddings, CNN/ RNN
12	Sep 27 (Tu)	Guest Lecture – Dr. Amitava Das: Glove, Word2Vec, Transformer Review: Reasoning for NLP
13	Sep 29 (Th)	Representation: Ontology, Knowledge Graph, QUIZ
14	Oct 4 (Tu)	Representation: Embeddings, Language Models
15	Oct 6 (Th)	Entity extraction
16	Oct 11 (Tu)	Guest Lecture – Dr. Amitava Das: Using lang models to solve NLP tasks

Announcements

GUEST LECTURES ON
LANGUAGE MODELS BY
DR. AMITAVA DAS

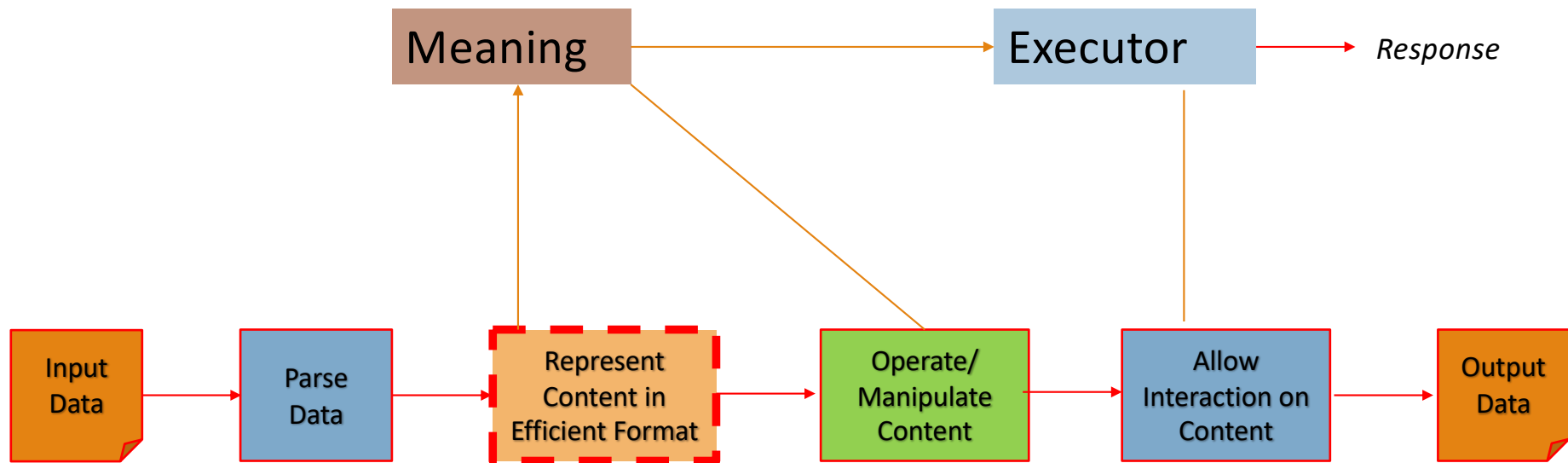
Project Report Format (A Starting Point*)

- Problem
- Related Work
- Data Sources
- Method / Solution Steps/ Algorithm
- Evaluation
- Significance of work / Discussion
- References

* Exact format depends on your project details. So, please change as appropriate.

Main Lecture

Semantics, Parsing and Representation



Semantics

- ***lexical semantics***: studies word meanings and word relations, and
- ***formal semantics***: studies the logical aspects of meaning, such as sense, reference, implication, and logical form
- ***conceptual semantics***: studies the cognitive structure of meaning

Source: Jurafsky & Martin,
Wikipedia (<https://en.wikipedia.org/wiki/Semantics>)

From Text to Meaning

- Shallow semantics
 - Input: text
 - Output: *lexical semantics*
- Deep semantics
 - Input: text
 - Output: *formal semantics*

Source: Abstract Meaning Representation for Sembanking,
<https://amr.isi.edu/a.pdf>

LOGIC format:

$\exists w, b, g:$
 $\text{instance}(w, \text{want-01}) \wedge \text{instance}(g, \text{go-01}) \wedge$
 $\text{instance}(b, \text{boy}) \wedge \text{arg0}(w, b) \wedge$
 $\text{arg1}(w, g) \wedge \text{arg0}(g, b)$

AMR format (based on PENMAN):

```
(w / want-01
 :arg0 (b / boy)
 :arg1 (g / go-01
       :arg0 b))
```

GRAPH format:

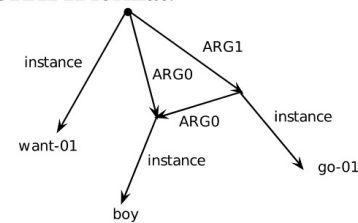


Figure 1: Equivalent formats for representing the meaning of “The boy wants to go”.

Review: Common Definitions

- **Corpus** (plural corpora): a computer-readable corpora collection of text or speech.
- **Lemma**: A lemma is a set of lexical forms having the same stem, the same major part-of-speech, and the same word sense. [Example: Cat and cats have same lemma.](#)
- **Word form**: The word form is the full inflected or derived form of the word. [Example: Cat and cats have different word forms.](#)
- **Word type**: Types are the number of distinct words in a corpus. if the set of words is V , the number of types is the word token vocabulary size $|V|$.
- **Word tokens**: The total number N of running words in the sentence / document of interest.
- **Code switching**: use multiple languages in a code switching single communicative act – [Example: Hindlish \(Hindi English\), Spanish \(Spanish English\)](#)

“They picnicked by [the](#) pool, then lay back on [the](#) grass and looked at [the](#) stars.”

- 16 tokens, 14 word types

Source: Jurafsky & Martin

Lexical Semantics

- Lemma
 - Sing, Mouse
- Word form
 - Sing, sang, sung
 - Mouse, mice
- Word sense
 - Mouse: a rodent
 - Mouse: an electronic pointing device

A lemma having many senses is called **Polysemous**

Synonymous and Similar Words

- **Synonym** - one word has a sense whose meaning is identical to a sense of another word
 - Two words are **synonymous** if they are **substitutable** one for the other in any sentence **without changing the truth conditions of the sentence, the situations in which the sentence would be true**
 - **Propositional meaning** – synonym words have the same propositional meaning (truth preserving)
- **Principle of contrast** – An assumption in linguistics is that difference in linguistic form (e.g., word form) is always associated with at least some difference in meaning
 - Water and H₂O are truth preserving but used in different context
 - Synonym words are used for approximate synonymy. Then, how similar are the words?

Source: Jurafsky & Martin

Word Similarity - SimLex-999

- Captures similarity between word pairs, mining the opinions of 500 annotators via Amazon Mechanical Turk on a scale of 1 to 10

Note: *similarity*, rather than *relatedness* or *association*

- Contains
 - 666 Noun-Noun pairs,
 - 222 Verb-Verb pairs
 - 111 Adjective-Adjective pairs

vanish	disappear	9.8
behave	obey	7.3
belief	impression	5.95
muscle	bone	3.65
modest	flexible	0.98
hole	agreement	0.3

Source: Jurafsky & Martin

- **Usage:** Evaluation of learning based approaches for finding word similarity by correlation

[SimLex-999: Evaluating Semantic Models with \(Genuine\) Similarity Estimation](#). 2014. Felix Hill, Roi Reichart and Anna Korhonen. *Computational Linguistics*. 2015
Website: <https://fh295.github.io/simlex.html>

Meaning (Semantics) versus Structure (Lexical)

Pair	Simlex-999 rating	WordSim-353 rating
<i>coast - shore</i>	9.00	9.10
<i>clothes - closet</i>	1.96	8.00

Example courtesy: <https://fh295.github.io/simlex.html>

Word Relatedness/ Association

- **Semantic Field:** related words from the same particular domain and bear structured relations with each other.
 - Example 1: cup, coffee
 - Example 2: scalpel, surgeon
 - Usually determined by experts in a field
- **Word Association Test/ Task:** how word meaning is stored in memory
 - Have people respond to word associations as a game; e.g., say the first word that comes to mind when one says “Doctor”
 - Applications
 - Used in marketing
 - Also evaluation of learning procedures discovering meaning (e.g., word embedding)

Source: Jurafsky & Martin

Sources:

- <https://psychology.jrank.org/pages/656/Word-Association-Test.html>,
- Establishing the Reliability of Word Association Data for Investigating Individual and Group Differences , Tess Fitzpatrick, David Playfoot, Alison Wray, Margaret J. Wright *Applied Linguistics*, Volume 36, Issue 1, February 2015, Pages 23–50, <https://doi.org/10.1093/applin/amt020>

Discovering Word Relatedness

- **Topic model:** a statistical notion of related words in a document. Hope is that meaningful topics will be from the same semantic field, but there is no guarantee
- Key idea
 - Topic: group of words
 - Counting words and grouping similar word patterns to infer topics within unstructured data.
 - Assumptions
 - Distributional hypothesis: similar topics make use of similar words
 - Statistical mixture hypothesis: documents talk about several topics
 - Perform unsupervised analysis/ clustering: given a corpus and number of topics (k), find k topics that are representative of key ideas in the corpus

References:

- Blog: <https://monkeylearn.com/blog/introduction-to-topic-modeling/>
- Tool: Gensim
- Paper: <https://www.jmlr.org/papers/volume3/blei03a/blei03a.pdf>

Frames, Slots: Frame Semantics

- Examples
 - "John sold a car to Mary"
 - "Mary bought a car from John"
 - "Mary paid John a undisclosed amount to get his car"
- To understand a word, one needs to understand the knowledge related to the word
 - In example: sell, buy, pay
- Capture knowledge in structures called **semantic frames** which have placeholders called slots (variables)
 - During parsing of sentences, values are filled
- **Frame semantics** is a theory of linguistic meaning developed by Charles J. Fillmore; related notion is semantic parsing

PropBank FrameSet

- A repository of formalized predicates
<https://proppbank.github.io/>

```
<roleset id="care.01"  
    name="having an opinion, feeling tenderly/strongly  
for/about">  
<roleset id="care.02"  
    name="liking/desiring/wanting">  
<roleset id="care.03"  
    name="tending, taking care of">  
<roleset id="care.04"  
    name="being cautious, taking care to">
```

Example: Care

<https://github.com/proppbank/proppbank-frames/blob/main/frames/care.xml>

Hindi – भेजा - *Beja*

Credits: <https://verbs.colorado.edu/propbank/framesets-hindi/Beja-v.html>

Example: Hindi Propbank

Roleset id: Beja.01 , to send, transport, ship something

Arg0: the one who sends something

Arg2: the recipient to whom something is sent

Arg1: the thing that is sent

Roleset id: Beja.02 , to send, transport, ship something

Arg0: the one who sends something

Arg2-gol: the place where something is sent

Arg1: the thing that is sent

Roleset id: Beja.03 , to make someone send something to someone

Argc: the causer- the one who makes someone send something

Arga: the intermediate causer

Arg0: the agent- the one who sends something

Arg2: the one to whom something is sent

Arg1: the thing that is sent

Roleset id: Beja.04 , to make someone send something to someplace

Argc: the causer- the one who makes someone send something

Arga: the intermediate causer

Arg0: the agent- the one who sends something

Arg2-gol: the place where something is sent

Arg1: the thing that is sent

Abstract Meaning Representation (AMR)

- Example: “The boy wants to go”
- AMR concepts are
 - English words (“boy”),
 - PropBank framesets (“want-01”), or
 - special key-words.
- Keywords include special entity types (“date-entity”, “world-region”, etc.), quantities (“monetary-quantity”, “distance-quantity”, etc.)
- logical conjunctions (“and”, etc).
- AMR uses approximately 100 relations

Source: Abstract Meaning Representation for Sembanking,
<https://amr.isi.edu/a.pdf>

LOGIC format:

$\exists w, b, g:$
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AMR format (based on PENMAN):

(w / want-01
:arg0 (b / boy)
:arg1 (g / go-01
:arg0 b))

GRAPH format:

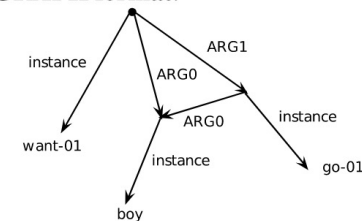


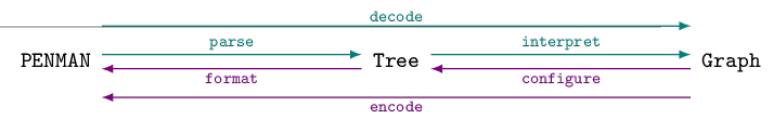
Figure 1: Equivalent formats for representing the meaning of “The boy wants to go”.

PENMAN Notation

```

; |----- Variable (this one is the graph's top)
; | |----- Instance relation
; | |-----
; | |-----
(d / drive-01
; |-----
; |----- Concept (node label)
; |----- Indicates the node's concept
; |----- Edge relation
; |-----
:ARG0 (h / he)
; |-----
; |----- Role (edge label)
:manner (c / care-04
; |----- Attribute relation
; |-----
:polarity -))
; |-----
; |----- Atom (or "constant")

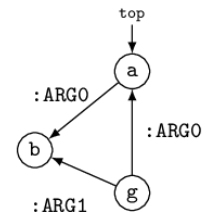
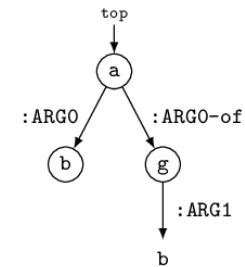
```



```

(a / alpha
 :ARG0 (b / beta)
 :ARG0-of (g / gamma
 :ARG1 b))

```



Credit: <https://penman.readthedocs.io/en/latest/structures.html>

Credit: <https://penman.readthedocs.io/en/latest/notation.html>

Sample Code – PENMAN/ AMR

Sample code:

<https://github.com/biplav-s/course-nl-f22/blob/main/sample-code/I9-semantics/PENMAN%20Notation%20-%20AMR.ipynb>

AMR Demo

<http://amparser.coli.uni-saarland.de:8080/>

AM Parser Demo

On this page, you can try out the AM Parser. This is a compositional neural parser which can parse English sentences into graph-based semantic representations. You can find more details in our ACL 2019 paper, or have a look at the source code on Github.

Sentence

The boy wants to go

Select graph formalisms into which the sentence will be parsed:

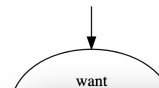
☒ DM ☒ PAS ☒ PSD ☒ EDS ☒ AMR-2017

Parse

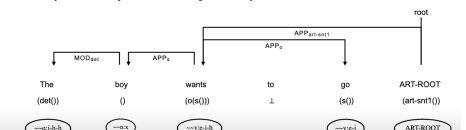
Parses

Parsing time: 3.2s, visualization time: 9.3s.

DM



DM (AM dependency tree)



Exercise: 5 mins

- Try your sentences online
- Look at output in different formats

Semantic Parsing

- Shallow semantic parsing
 - Also called: slot-filling or frame semantic parsing
 - "show me flights from **Boston** to **Dallas**"
- Deep semantic parsing
 - "show me flights from **Boston** to **anywhere** that has flights to **Dallas**"
 - Reference to quantifiers

Applications

- Paraphrasing
- Machine comprehension
- Question-answering
- Dialog

References:

- ACL 2020 Tutorial on Semantic Parsing
- https://en.wikipedia.org/wiki/Semantic_parsing

Semantic Parsing

Language to Meaning



Example Task

Database Query

What states
border Texas?



Oklahoma
New Mexico
Arkansas
Louisiana

Source:
ACL 2020 Tutorial on Semantic Parsing

Resources: Semantic Parsing Libraries

- Open Sesame
 - Given English sentence, predicts FrameNet frames
 - <https://github.com/swabhs/open-sesame>
- AMRLib
 - Python library for AMR parsing, generation and visualization simple
 - <https://github.com/bjascob/amrlib>

Review: Lexical Meaning – Common Terms

- **Synonym:** same/ similar meaning
 - start-begin, finish-end, far-distant
- **Antonym:** opposite meaning
 - Far – near, clever - stupid, high - low, big – small
- **Homonym:** identical in spelling and pronunciation
 - bear, bank, ...
- **Homophones:** sounds identical but are written differently
 - site-sight, piece-peace.
- **Homograph:** written identically but sound differently
 - Potato, tomato, lead, wind, minute
- **Polysemy:** a word or phrase which has two (or more) different meanings (i.e., senses)
 - Duck, sharp

Source: Mausam

More Terms

- **Affective meanings** or **connotation**: word's meaning that are related to a writer or reader's emotions, sentiment, opinions, or evaluations
 - Positive evaluation: good, happy
 - Negative evaluation:
- **Sentiment**: Positive or negative evaluation expressed through language
 - Scherer's Typology of Affective States

Source: Jurafsky & Martin

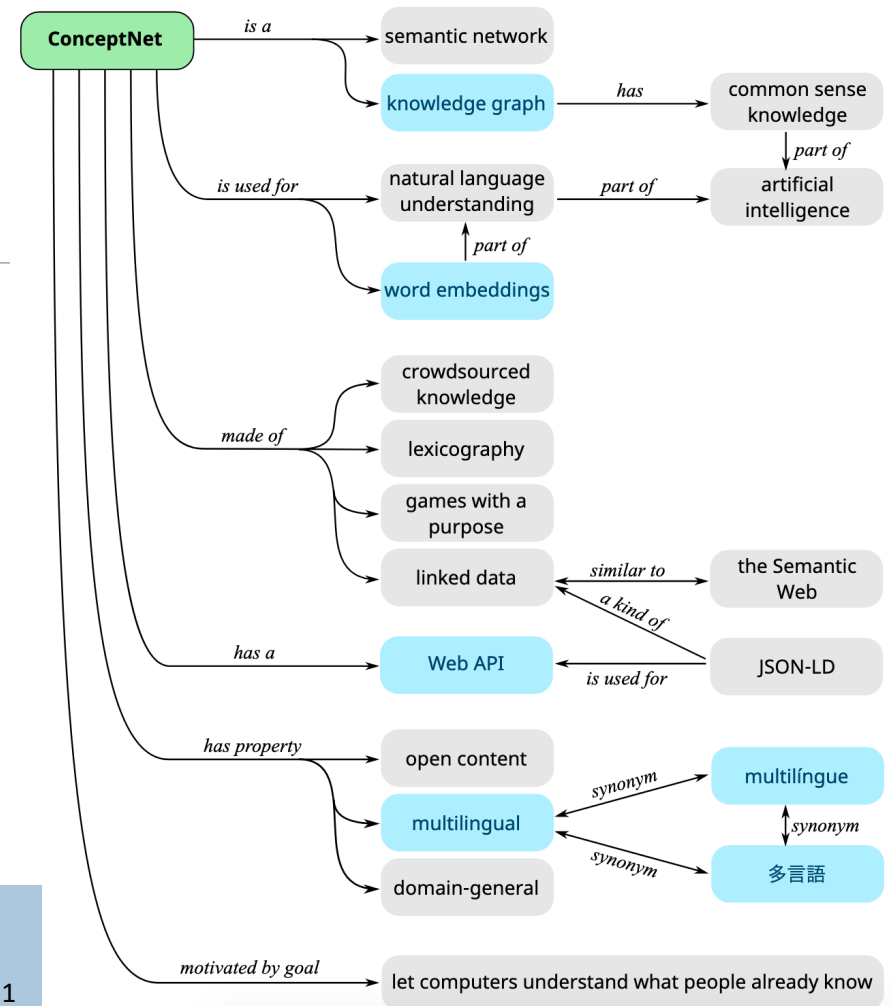
ConceptNet

- NLP focused graph knowledge graph that connects words and phrases of natural language with labeled edges.
- Concepts collected from experts, crowd-sourcing, and games with a purpose
- Supports multiple languages
- Provides "loose" semantics - relatedness

Details: <http://conceptnet.io/>,

<https://github.com/commonsense/conceptnet5/wiki>,

Paper: <https://www.aaai.org/ocs/index.php/AAAI/AAAI17/paper/viewFile/14972/14051>



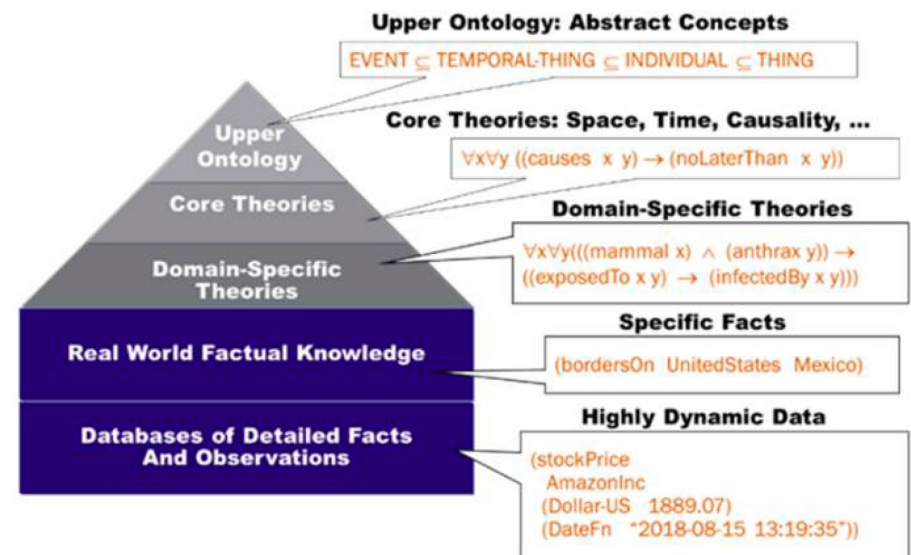
Demonstration - ConceptNet

Examples:

- Concepts:
 - Word: <http://conceptnet.io/c/en/word> ,
 - duck: <http://conceptnet.io/c/en/duck>
- Relationships:
 - <http://conceptnet.io/s/resource/wordnet/rdf/3.1>

Project CYC

- A large ontology to capture the world and human common sense
 - Doug Lenat lead team of computer scientists, computational linguists, philosophers, and logicians
 - Identify and formally axiomatize the tens of millions of rules about world
 - 35+ years effort by Cycorp
- Reasoners on the ontology to make decisions
 - 1000+ specialized reasoners



Details: <https://www.cyc.com/>

Source: Cyc White Paper

Cyc Details

- Ontology of about 1.5 million general concepts (e.g., taxonomically “placing” terms like eyes, sleep, night, person, unhappiness, hours, posture, being woken up, etc.);
- More than 25 million general rules and assertions involving those concepts
 - *“Most people sleep at night, for several hours at a time, lying down, with their eyes closed, they can be awakened by a loud noise but don’t like that, “*
- Domain-specific extensions to the common sense ontology and knowledge base
 - healthcare, intelligence, defense, energy, transportation and financial services.
- Promoting synergistic use of ontology and learning based approaches (now)

Source: White Paper – Cyc Technology Overview

Lecture 9: Concluding Comments

- We reviewed how to give semantics to words and documents
- Can be human supervised or learning based or combined
- Can be generic or task-oriented

Concluding Segment

About Next Lecture – Lecture 10

Lecture 10 Outline

- Machine Learning for NLP
 - Supervised learning
 - Unsupervised learning
 - Neural methods