

Introduction to NLP

Semantic Analysis

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Learning Outcomes of This Lecture

Overview of main tasks in semantic analysis of language
Understand how to develop a Word Sense Disambiguation approach
Insight into Semantic Role Labeling
Insight into Coreference Resolution



Overview

Semantic Analysis
Word Sense Disambiguation
Semantic Role Labelling
Coreference Resolution



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Semantic Analysis

Semantic analysis of language data is concerned with how words refer to concepts about objects in the world

Challenge is **ambiguity**: which word(s) refer to which concept(s)?



Levels of Semantic Analysis

Lexical Semantics

meaning of words

Compositional Semantics

combined meaning of words in a phrase or sentence

Discourse Semantics

combined meaning of words across several sentences



Semantic Analysis Tasks

Lexical Semantic Analysis: word sense disambiguation identify correct word meaning or 'word sense'

Compositional Semantic Analysis: **semantic role labeling** identify correct 'semantic role' for each word/phrase

Discourse Semantic Analysis: **coreference resolution**identify which words/phrases refer to the same 'entity' across sentences



Overview

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Coreference Resolution



Lexical Semantic Ambiguity

Words can have different meanings due to homonymy



table in the interpretation of 'furniture'

table in the interpretation of 'accounting'





Word Sense Disambiguation

Identify correct word meaning or 'word sense' for ambiguous words:

To add additional columns, drag the field to the [table ACCOUNTING] until you see an insertion point.

Whether on the mantelpiece or on the dining [table FURNITURE], vases are a beautiful addition to any home.



Word Senses

Word senses are defined by a **semantic lexicon** such as WordNet A semantic lexicon defines word senses **explicitly**, by use of lists Word senses can also be defined **implicitly** by use of clusters **Here we focus on explicitly defined word senses**



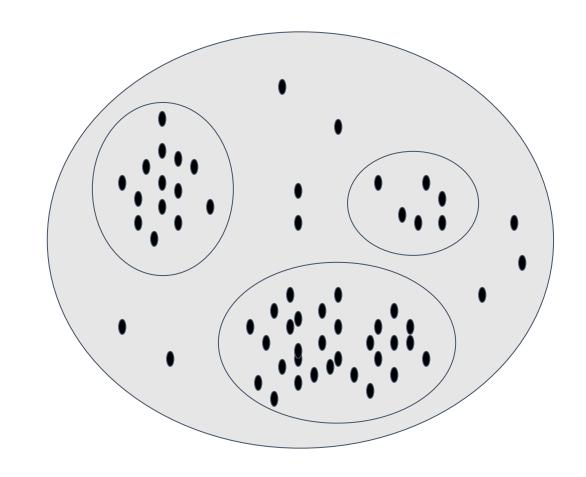
Explicit vs. Implicit Word Senses

table

Sense 1: FURNITURE

Sense 2: ACCOUNTING

Sense 3: GEOGRAPHICAL





WordNet Senses for table

Remember that WordNet uses 'synsets' to express word senses

- S: (n) table#1, tabular array#1 (a set of data arranged in rows and columns) "see table 1"
- S: (n) table#2 (a piece of furniture having a smooth flat top that is usually supported by one or more vertical legs) "it was a sturdy table"
- S: (n) table#3 (a piece of furniture with tableware for a meal laid out on it) "I reserved a table at my favorite restaurant"
- S: (n) mesa#1, table#4 (flat tableland with steep edges) "the tribe was relatively safe on the mesa but they had to descend into the valley for water"
- S: (n) table#5 (a company of people assembled at a table for a meal or game) "he
 entertained the whole table with his witty remarks"
- S: (n) board#4, table#6 (food or meals in general) "she sets a fine table"; "room and board"



Data Labeling - WordNet Glosses

Word	Sense	SynSets	WordNet 'glosses' (examples)
table	#1	table#1, tabular array#1	see table 1
	#2	table#2	it was a sturdy table
	#3	table#3	I reserved a table at my favourite restaurant
	#4	table#4, mesa#1	the tribe was relatively safe on the table
	#5	table#5	he entertained the whole table
	#6	table#6, board#4	she sets a fine table



Data Labeling - Corpus Data

Word	Sense	Corpus Data (manual annotation)
table	#1	To add additional columns, drag the field to the table until you see an insertion point.
		In addition to rows, you can also divide a table into columns.
	#2	Whether on the mantelpiece or centre stage on the dining table , vases are a beautiful addition to any home.
		Our dining tables and chairs are of the highest quality, crafted from solid premium wood.



Basic Algorithm for WSD - 'Simplified Lesk'

function SIMPLIFIED LESK(word, sentence) **returns** best sense of word

```
best-sense ← most frequent sense for word

max-overlap ← 0

context ← set of words in sentence

for each sense in senses of word do

signature ← set of words in the gloss and examples of sense

overlap ← COMPUTEOVERLAP(signature, context)

if overlap > max-overlap then

max-overlap ← overlap

best-sense ← sense

end

return(best-sense)
```

Jurafsky & Martin p365



Lesk Algorithm – Example WSD of table

The toolbar includes a basic **table** drawing feature that lets you divide the canvas into rows and columns that you can use to organize your text.



Lesk Algorithm – WordNet Senses

The toolbar includes a basic **table** drawing feature that lets you divide the canvas into rows and columns that you can use to organize your text.

Word	Sense	SynSets
table	#1	table#1, tabular array#1
	#2	table#2



Lesk Algorithm – Signature

The toolbar includes a basic **table** drawing feature that lets you divide the canvas into rows and columns that you can use to organize your text.

Word	Sense	Signature (from corpus data)
table	#1	add additional columns divide drag field insertion point rows see
	#2	beautiful centre chairs crafted dining highest home mantelpiece premium quality solid stage sturdy vases wood



Lesk Algorithm – Overlap

The toolbar includes a basic **table** drawing feature that lets you <u>divide</u> the canvas into rows and <u>columns</u> that you can use to organize your text.

Word	Sense	Signature
table	#1	add additional <u>columns</u> <u>divide</u> drag field insertion point rows see
	#2	beautiful centre chairs crafted dining highest home mantelpiece premium quality solid stage sturdy vases wood



Lesk Algorithm – Best Sense

The toolbar includes a basic **table** drawing feature that lets you <u>divide</u> the canvas into rows and <u>columns</u> that you can use to organize your text.

Word	Sense	Signature
table	#1	add additional columns divide drag field insertion point rows see
	#2	beautiful centre chairs crafted dining highest home mantelpiece premium quality solid stage sturdy vases wood



WSD - Vector Space Model

WSD target:

To add columns and rows, drag the field to the table until you see an insertion point.

Corpus data for senses:

In addition to rows, you can also divide a table#1 into columns.

This table#2 is of the highest quality, crafted from solid wood.

Vocabulary (nouns in WSD target and corpus data):

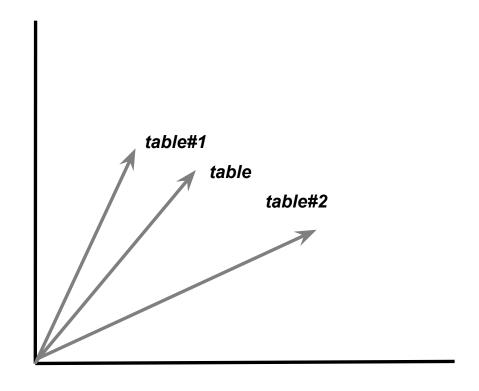
[columns, rows, field, point, quality, wood]

Vectors (binary):

table [1,1,1,1,0,0] table#1 [1,1,0,0,0,0] table#2 [0,0,0,0,1,1]



WSD - Vector Space Visualization





WSD Training

Supervised

WordNet glosses

Manual annotation of corpus data with word senses (see link below)

Semi-supervised

'Distant supervision' on Wikipedia categories

Unsupervised

Graph-based WSD



Wikipedia Categories - table

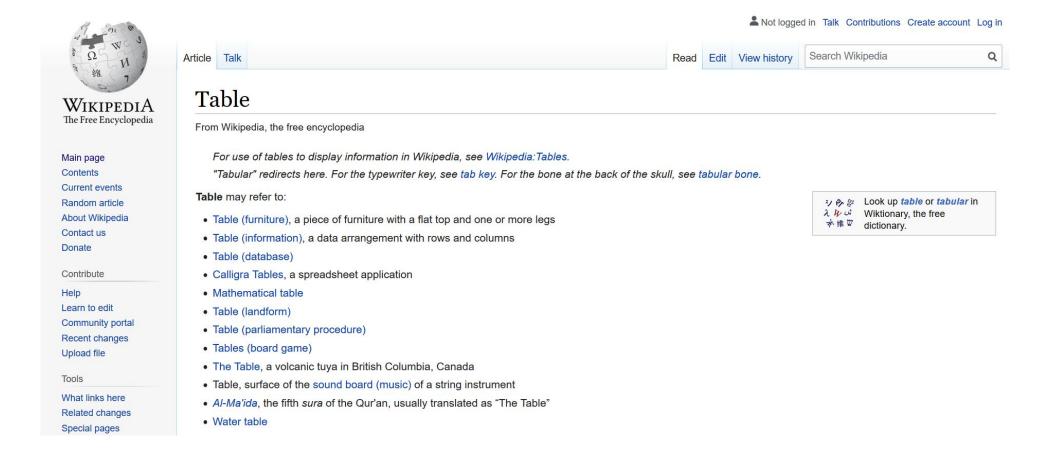


table as "furniture, information, landform"

https://en.wikipedia.org/wiki/Table (furniture)

A **table** is an item of furniture with a flat top and one or more legs, used as a surface for working at, eating from or on which to place things. ... There are also a range of specialized types of **tables**, such as drafting **tables**, used for doing architectural drawings, and sewing **tables**.

https://en.wikipedia.org/wiki/Table (information)

A **table** is an arrangement of data in rows and columns, or possibly in a more complex structure. **... Tables** appear in print media, handwritten notes, computer software, architectural ornamentation, traffic signs, and many other places. ... In books and technical articles, **tables** are typically presented apart from the main text in numbered and captioned floating blocks.

https://en.wikipedia.org/wiki/Table (landform)

The term "flat" is relative when speaking of **tables**, and often the naming or identification of a **table** (or **table** hill or mountain) is based on the appearance of the terrain feature from a distance or from below it.



Labelled Data by Wikipedia Category

https://en.wikipedia.org/wiki/Table (furniture)

A **table:furniture** is an item of furniture with a flat top and one or more legs, used as a surface for working at, eating from or on which to place things. ... There are also a range of specialized types of **tables:furniture**, such as drafting **tables:furniture**, used for doing architectural drawings, and sewing **tables:furniture**.

https://en.wikipedia.org/wiki/Table (information)

A **table:information** is an arrangement of data in rows and columns, or possibly in a more complex structure. ... **Tables:information** appear in print media, handwritten notes, computer software, architectural ornamentation, traffic signs, and many other places. ... In books and technical articles, **tables:information** are typically presented apart from the main text in numbered and captioned floating blocks.

https://en.wikipedia.org/wiki/Table (landform)

The term "flat" is relative when speaking of **tables:landform**, and often the naming or identification of a **table:landform** (or **table:landform** hill or mountain) is based on the appearance of the terrain feature from a distance or from below it.



Aligning Wikipedia & WordNet for WSD

Wikipedia: Table_(furniture) - WordNet table#2

A **table#2** is an item of furniture with a flat top and one or more legs, used as a surface for working at, eating from or on which to place things.

Wikipedia: Table_(information) - WordNet table#1

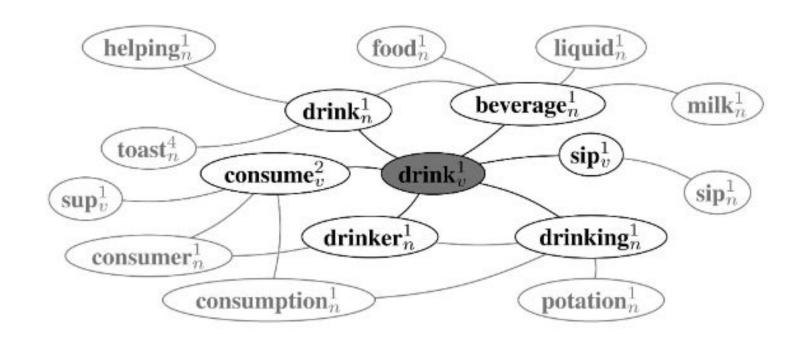
A table#1 is an arrangement of data in rows and columns, or possibly in a more complex structure.

Wikipedia: Table_(landform) - WordNet table#4

The term "flat" is relative when speaking of **tables#4**, and often the naming or identification of a **table#4** (or **table#4** hill or mountain) is based on the appearance of the terrain feature from a distance or from below it.

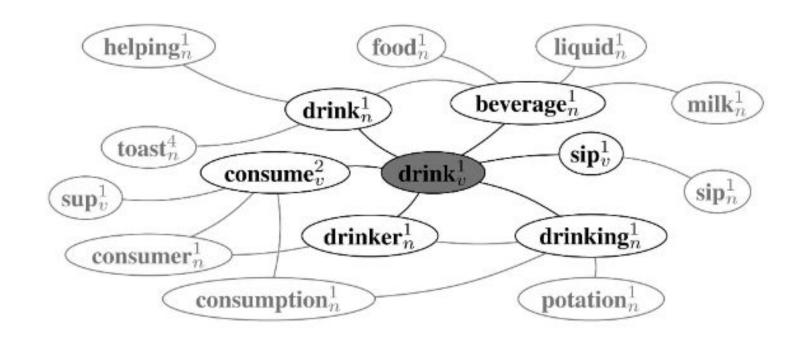


WordNet graph for the verb *drink* with word sense: *drink#1, imbibe#3 - take in liquids*



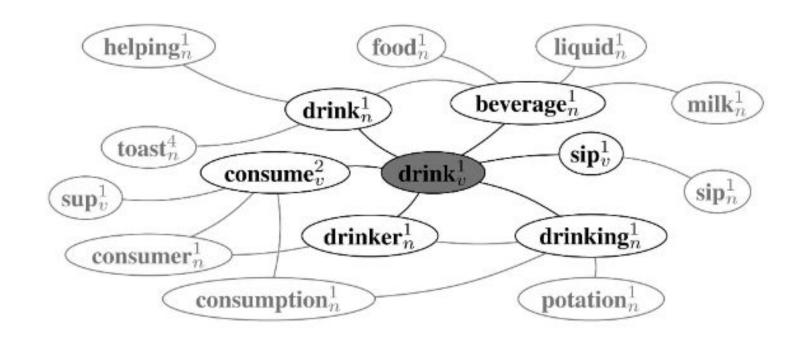


Can we use this graph to disambiguate words in the sentence "She drank some milk."

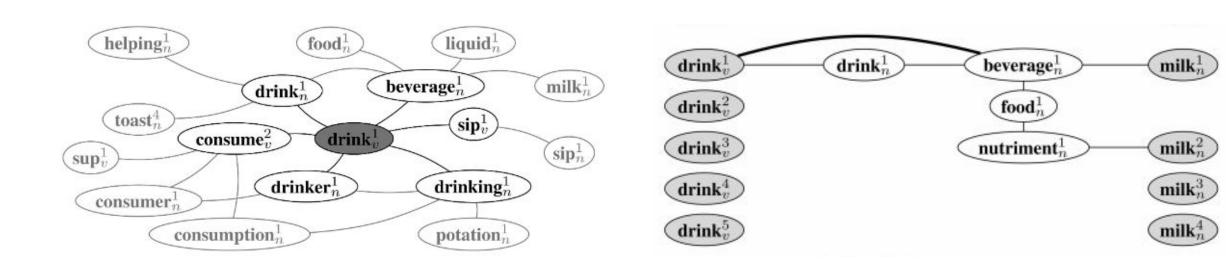




Identify shortest path between "drank" (verb "to drink") and "milk" (noun)

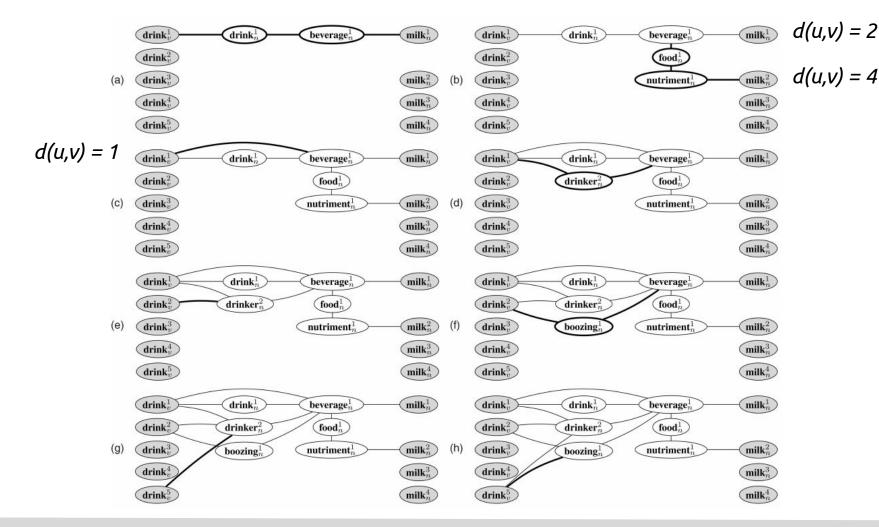






d(u,v)=1

"She drank some milk."





Overview

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Coreference Resolution



Entities, Roles & Events

Consider the following sentence

"Kristing hit Scott with a baseball."

There are several actors/items ('entities') with different 'roles' in the hitting 'event'

Kristina

Scott

baseball



Entities, Roles & Events

Different sentences with 'entities' in the same 'roles' can express the same 'event'

Kristing hit Scott with a baseball.

Scott was hit by Kristina with a baseball.

Scott was hit with a baseball by Kristina.



Semantic Roles

We can define 'semantic roles' for each entity of an event expressed by a sentence, e.g.

AGENT (AGT), EXPERIENCER (EXP), INSTRUMENT (INS)

[Kristina AGT] hit [Scott EXP] [with a baseball INS].

[Scott EXP] was hit by [Kristina AGT] [with a baseball INS].

[Scott EXP] was hit [with a baseball INS] by [Kristina AGT].



Semantic Roles - Overview

AGENT The volitional causer of an event

EXPERIENCER The experiencer of an event

FORCE The non-volitional causer of the event

THEME The participant most directly affected by an event

RESULT The end product of an event

CONTENT The proposition or content of a propositional event

INSTRUMENT An instrument used in an event

BENEFICIARY The beneficiary of an event

SOURCE The origin of the object of a transfer event

GOAL The destination of an object of a transfer event

The waiter spilled the soup.

John has a headache.

The wind blows debris from the mall into our yards.

Only after Benjamin Franklin broke the ice...

The city built a regulation-size baseball diamond...

Mona asked "You met Mary Ann at a supermarket?"

He poached catfish, stunning them with a shocking device...

Whenever Ann Callahan makes hotel reservations for her boss...

I flew in from Boston.

I drove to Portland.

Note: semantic roles are also referred to as 'thematic roles'

Jurafsky & Martin p374



Semantic Roles - Use

Semantic roles enable 'event identification' across different sentences, which can be very useful in applications such as Question Answering, e.g.

Question: "Which company bought Google last year?"

Answer Candidates: "Fitbit bought by Google in 2019"

"Last year it was Fitbit that Google bought"



Semantic Roles - Challenges

Semantic roles are not very well defined with different names and interpretations of roles across the literature

Automatic 'semantic role labeling' depends largely on supervised training with limited labeled data existing



Semantic Role Labeling

Identification of semantic roles of each 'argument' of each 'predicate' in a sentence Basic algorithm for Semantic Role Labeling (SRL):

function SEMANTICROLELABEL(words) returns labeled tree

```
parse ← PARSE(words)

for each predicate in parse do

for each node in parse do

featurevector ← EXTRACTFEATURES(node, predicate, parse)

CLASSIFYNODE(node, featurevector, parse)
```

Jurafsky & Martin Ch20



Semantic Role Labeling - Feature Extraction

[AGENT The chancellor] announced [THEME new policy measures].

Predicate: announce

Voice (whether predicate is active or passive): active

Position (before/after) of classified constituent (phrase) with regard to predicate: **before**

Head Word of classified constituent (phrase): chancellor



Semantic Role Labeling - Feature Extraction

[AGENT The chancellor] announced [THEME new policy measures].

Phrase Type (NP, PP, ...) of classified constituent (phrase):

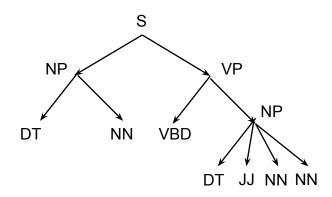
Minimal Path from classified constituent (phrase) to predicate:

Sub-categorization (phrase structure rule) for parent of predicate:

NP

NP↑**S**↓**VP**↓**VBD**

VP→VBD,NP



Semantic Role Labeling - Feature Extraction

[AGENT The chancellor] announced [THEME new policy measures].

Predicate: announce

Voice (whether predicate is active or passive): active

Phrase Type (NP, PP, ...) of classified constituent:

Minimal Path from classified constituent to predicate: NP↑S↓VP↓VBD

Position (before/after) of classified constituent with regard to predicate: **before**

Head Word of classified constituent: chancellor

Sub-categorization for parent of predicate: VP→VBD,NP

AGENT [announce, active, NP, NP \uparrow S \downarrow VP \downarrow VBD, before, chancellor, VP \rightarrow VBD,NP]



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Coreference Resolution

Identify words and/or phrases referring to the same entity

Application of coreference resolution for example in dialog systems:

USER: "How late is the 404 bus to Dangan?"

SYSTEM: "The 404 bus to Dangan leaves at 12:35"

USER: "Does it stop at the university?" it: bus

SYSTEM: "The bus to Dangan stops at the university"

USER: "When will it stop there?" it: bus

there: the university



Coreference Resolution - Example



Coreference Resolution - Mentions

Identify all mentions - coreference candidates



Coreference Resolution - Mentions

Mentions referring to Barack Obama



Coreference Resolution - Mentions

Mentions referring to *Hillary Rodham Clinton*



Mention Detection and Mention Clustering

Mention Detection

"Barack Obama nominated Hillary Rodham Clinton as his secretary of state on Monday. He chose her because she had foreign affairs experience as a former First Lady."

Mention Clustering



Mention Detection - Kinds of Mentions

Mention is a span of text referring to an entity

There are three kinds of mentions:

- **Pronouns**: *I, your, it, she, him, ...*
- Named Entities: person, location, organization, ...
- Noun Phrases: "a dog," "the big fluffy cat stuck in the tree"



Mention Detection - NLP Tools to Use

For detection of each kind of mention use:

- Part-of-speech tagger to identify pronouns
- Named Entity Recognizer (NER) system to identify named entities
- Constituency (Phrase) parser to identify noun phrases



Mention Clustering - Features

Recency - more recently mentioned entities are preferred

John went to a movie. Jack went as well. He was not busy.

He = Jack (He ≠ John)

Syntactic agreement

Jack gave Mary a gift. She was excited.

She = Mary (She \neq Jack)

Other syntactic constraints

John bought him a new car.

him ≠ John

Semantic compatibility

... the mining **conglomerate** ... the **company** ...

company =

conglomerate

Grammatical Role - prefer entities in the subject position

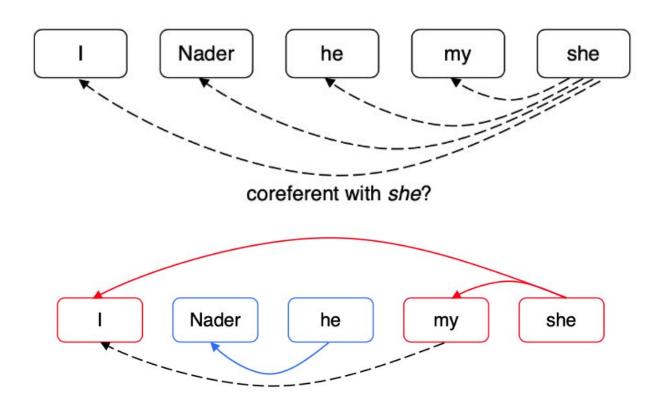
John went to a movie with **Jack**. **He** was not busy.

He = Jack (He ≠ John)



Mention Clustering - Example

"I voted for Nader because he was most aligned with my values," she said.





Lab of this Week

Word Sense Disambiguation exercises with WordNet

