# Introduction to Natural Language Processing

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### Objective of the course

- Understanding of basics of Natural Language
   Processing
- Understanding Natural Language Processing using Python programming



#### Topics which will be covered

- Tokenization, Morphology, Stemming
- Regular expression
- POS Tagging
- Language Modeling
- Parsing
- Semantics, Representing and Understanding Meaning,
   Knowledge Representation
- NLP topics: Tex Summarization, Word Sense Disambiguation, Information Extraction, Sentiment Analysis,



#### Reference materials

- Main Text(s):
  - Natural Language Understanding: James Allan
  - Speech and NLP: Jurafsky and Martin
  - Foundations of Statistical NLP: Manning and Schutze
- Other References:
  - NLP a Paninian Perspective: Bharati, Cahitanya and Sangal
  - Statistical NLP: Charniak
- Journals
  - Computational Linguistics, Natural Language Engineering, AI, AI Magazine, IEEE SMC
- Conferences
  - ACL, EACL, COLING, MT Summit, EMNLP, IJCNLP, HLT, ICON, SIGIR, WWW, ICML, ECML



#### What is NLP?

- Branch of Al
- To give computers the ability to process human language
- 2 Goals
  - Science Goal: Understand the way language operates
  - Engineering Goal: Build systems that analyze and generate language; reduce the man machine gap



## **Ambiguity**

This is what makes NLP challenging

The crux of the problem



### Stages of processing

- Phonetics and phonology
- Morphology
- Lexical Analysis
- Syntactic Analysis
- Semantic Analysis
- Pragmatics
- Discourse



#### **Phonetics**

- Processing of speech
- Challenges
  - Homophones: bank (finance) vs. bank (river bank)
  - Near Homophones: maatraa vs. maatra (hin)
  - Word Boundary
    - aajaayenge (aa jaayenge (will come) or aaj aayenge (will come today)
    - I got [ua]plate
  - Disfluency: ah, um, ahem etc.



## Morphology

- Word formation rules from root words
- Nouns: Plural (boy-boys); Gender marking (czar-czarina)
- Verbs: Tense (*stretch-stretched*); Aspect (*e.g. perfective sit-had sat*); Modality (e.g. *request khaanaa* → *khaaiie*)
- First crucial first step in NLP
- Languages rich in morphology: e.g., Dravidian, Hungarian, Turkish
- Languages poor in morphology: Chinese, English
- Languages with rich morphology have the advantage of easier processing at higher stages of processing



### Lexical Analysis

 Essentially refers to dictionary access and obtaining the properties of the word

```
e.g. dog
noun (lexical property)
take-'s'-in-plural (morph property)
animate (semantic property)
4-legged (-do-)
carnivore (-do)
```

Challenge: Lexical or word sense disambiguation



### Lexical Disambiguation

#### First step: part of Speech Disambiguation

- Dog as a noun (animal)
- Dog as a verb (to pursue)

#### Sense Disambiguation

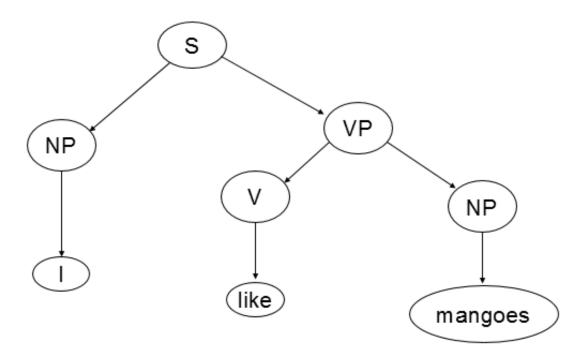
- Dog (as animal)
- Dog (as a very detestable person)



## **Syntax Processing Stage**

• Structure Detection:

#### I like mangoes





### Syntax Processing Stage contd...

- Scope
  - 1.The old men and women were taken to safe locations
  - (old men and women) vs. ((old men) and women)
  - 2. No smoking areas will allow Hookas inside
- Preposition Phrase Attachment
  - I saw the boy with a telescope (who has the telescope?)
  - I saw the boy with the pony-tail (world knowledge: pony-tail cannot be an instrument of seeing)



### Semantic Analysis

- Representation in terms of
  - Predicate calculus/Semantic Nets/Frames/Conceptual Dependencies and Scripts
- John gave a book to Mary
  - Give action: Agent: John, Object: Book, Recipient: Mary
- Challenge: ambiguity in semantic role labeling
  - (Hin) aapko mujhe mithaai khilaanii padegii (ambiguous in Marathi and Bengali too; not in Dravidian languages)



#### **Pragmatics**

- Very hard problem
- Model user intention
  - Tourist (in a hurry, checking out of the hotel, motioning to the service boy): Boy, go upstairs and see if my sandals are under the divan. Do not be late. I just have 15 minutes to catch the train.
  - Boy (running upstairs and coming back panting): yes sir, they are there.



#### Discourse

- Processing of sequence of sentences
- Mother to John:

John go to school. It is open today. Should you bunk? Father will be very angry.

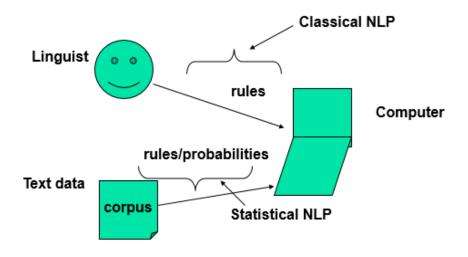
- Ambiguity of open
- bunk what?
- Why will the father be angry?
  - Complex chain of reasoning and application of world knowledge

```
Ambiguity of father
father as parent
or
father as headmaster
```



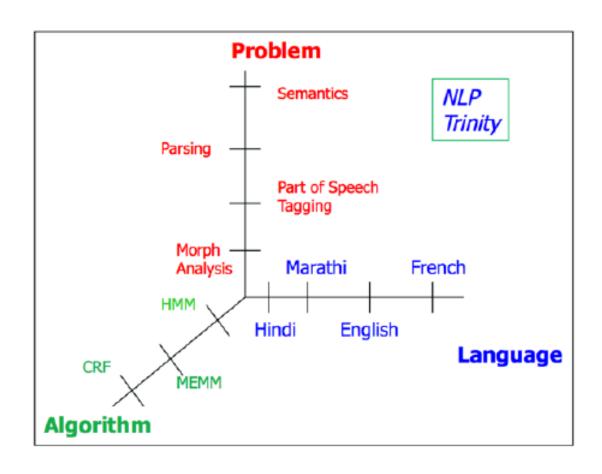
# Two Views of NLP and the Associated Challenges

- Classical View:
  - Layered processing; Ambiguities
- Statistical/Machine Learning View:
  - Prediction



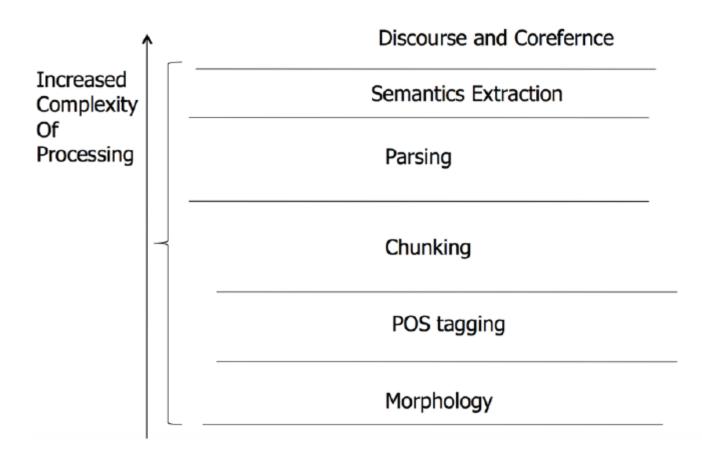


## **NLP Trinity**





## **NLP Layers**





### **NLP Techniques**

- Rule based approach
- Knowledge based approach
- Machine Learning based approach
  - Supervised approach
  - Unsupervised approach
  - Semi-supervised approach



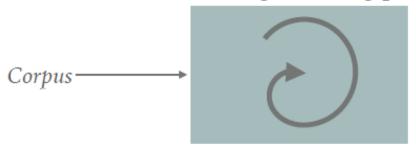
## What is the output of NLP system?

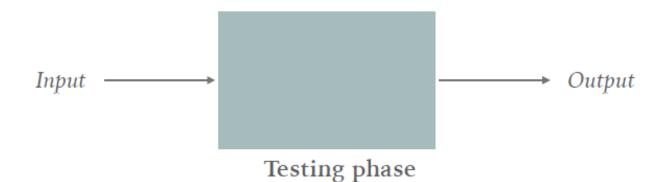
- Option 1: A set of rules
- Option 2: A set of probability values



## Basic building blocks

Learning/Training phase





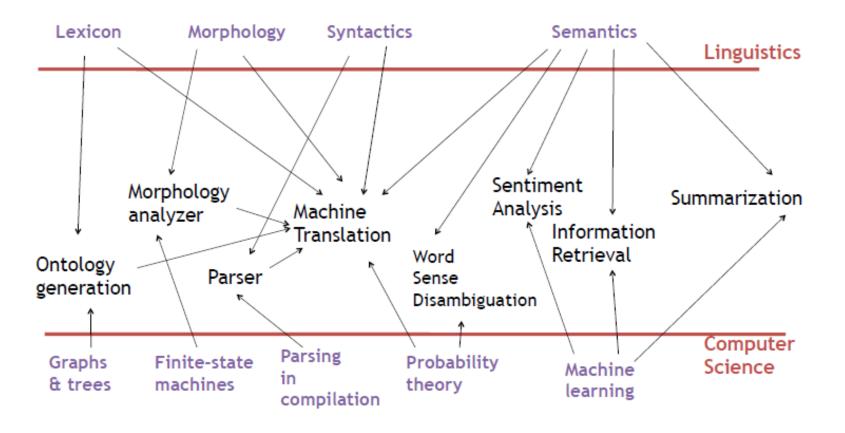


## Example: Parts-Of-Speech tagging

Labeled dataset of Learning/Training phase sentences e.g. She PRONOUN dances VERB e.g. Dances NOUN are VERB good\_ADJECTIVE e.g. He PRONOUN is VERB A ARTICLE singer NOUN Input Output A ARTICLE A singer dances singer NOUN Testing phase dances VERB



# NLP: At the confluence of Linguistics & computer science





#### Lexical Resources

- Backbone of every NLP tasks
- Computer do not understand natural languages, like English, Hindi, etc.
- We need to train them based on available resources



### Corpus

- A collection of text, it can be a plain text document containing data or a structured XML file containing data along with its descriptors and headers
- Corpora Plural form of corpus
- Examples:
  - Monolingual corpora
  - Comparable
  - Parallel
  - Annotated corpora: pos-tagged, sense-tagged



#### Corpus contd...

Monolingual corpus

A boy is sitting in the kitchen
A boy is playing a tennis
Some men are watching tennis
A girl is holding a black book



### Corpus contd...

- Comparable corpus
  - A collection of "similar" texts in different languages or in different varieties of a language.
  - A pair of corpora in two different languages, which come from the same domain.
  - An example would be a corpus of articles about football from English and Danish newspapers; or legal contracts in Spanish and Greek.



### Corpus contd...

• Parallel corpus: the texts in one language (L1) are translations of texts in the other language (L2).

L1 (English)	L2 (Hindi)
A boy is sitting in the kitchen	एक लड़का रसोई में बैठा है
A boy is playing a tennis	एक लड़का एक टेनिस खेल रहा है
Some men are watching tennis	कुछ पुरुष टेनिस देख रहे हैं
A girl is holding a black book	एक लड़की एक काली पुस्तक पकड़े हुए है

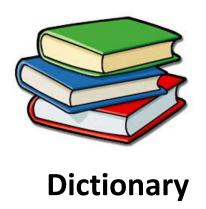


#### Corpus contd..

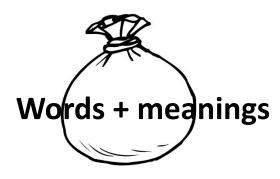
- Annotated corpus:
  - Corpus annotated with different label to each unit in the corpus
- Pos-tagged corpus
  - She\_PRON eats\_Verb rice\_NOUN
- Sense-tagged corpus
  - The city\_18406385 is\_22579744 famous\_41426596 for its majestic\_41333338 forts\_13350250, palaces\_13834381 which attract\_21492358 tourists\_110557758 from the world\_19138104.

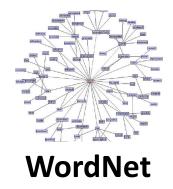


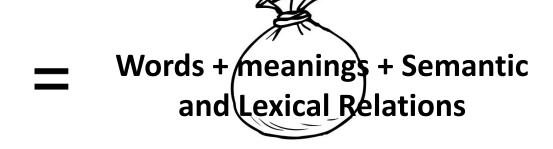
#### What is WordNet?













#### What is WordNet? Contd...

- A lexical knowledge database for a language
- Consists of synsets and lexico-semantic relations
- Categorizes synsets into four main parts-of-speech categories: nouns, adjectives, adverbs and verbs
- Monolingual WordNet
  - English, Hindi, Sanskrit
- Multilingual WordNet
  - IndoWordNet, EuroWordNet, BabelNet



### WordNet Synset

- Each synset consist of:
  - Sense ID
  - Parts-of-speech category
  - Synset Members (Synonyms words)
  - Gloss or Concept Definition
  - Example Sentence
- Synset of a boy:

(10305010) (n) male child, boy (a youthful male person) "the baby was a boy"; "she made the boy brush his teeth every night"; "most soldiers are only boys in uniform"

## WordNet Lexico-Semantic Relations

- Synonymy
- Antonymy
- Gradation
- Hypernymy / Hyponymy
- Meronymy / Holonymy
- Entailment
- Attribute
- Nominalization
- Ability Link
- Capability Link
- Function Link



#### **Lexical Relations**

- Relation between words
- Synonymy: relationship between words in a synset.
  - {plant, flora}, 'plant' and 'flora' are related through synonymy relation.
- Antonymy: relationship between words having an opposite meaning.
  - 'day' and 'night' are antonyms of each other.
- Gradation:
  - 'morning', 'afternoon', 'evening' are related through gradation relation



#### Semantic Relations

- Relation between synsets
- Hypernymy / Hyponymy: is-a-kind-of relation
  - 'fruit' is a hypernym of 'mango' and 'mango' is a hyponym of 'fruit'.
- Meronymy / Holonymy: part-whole relation
  - 'hand' is a meronym of 'body' and 'body' is a holonym of 'hand'



### Semantic Relations contd...

- Fntailment:
  - 'snore' entails 'sleep'
- Attribute: relationship between noun and adjective synsets
  - 'hot' is a value of or attribute of 'temperature'
- Nominalization: relationship between noun and verb synsets
  - 'service' nominalizes the verb 'serve'

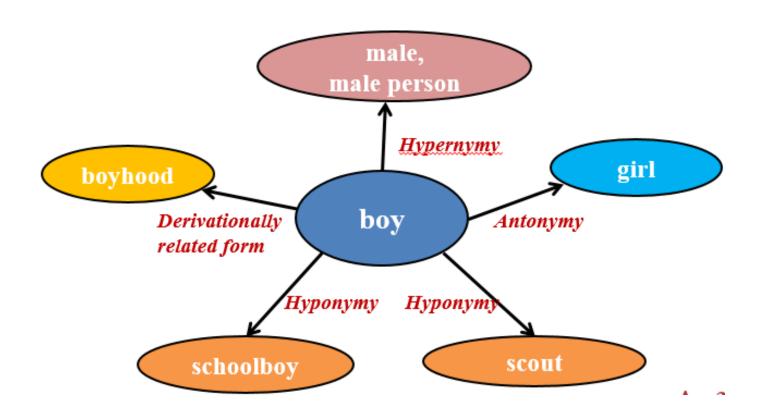


### Semantic Relations contd...

- Ability Link: specifies the inherited features of a nominal concept
  - 'animal' and 'walk', 'fish' and 'swim'
- Capability Link: relationship specifies the acquired features of a nominal concept
  - 'person' and 'swim'
- Function Link: relationship specifies the function of a nominal concept
  - 'vehicle' and 'move' and 'teacher' and 'teach'



# WordNet Lexico-Semantic Relations





### Some important wordnets

- English WordNet (Fellbaum, 1998):
  - First semantic net created at Princeton University
- Hindi WordNet (Narayan et. al, 2002)
  - First Indian language Wordnet which is created from English WordNet using expansion approach at IIT Bombay
- IndoWordnet (Bhattacharyya, 2010)
  - A Multilingual Wordnet for 17 Indian Languages
- BabelNet (Navigli, 2010)
  - A very large, wide coverage multilingual semantic network
  - 271 languages, 14 million synsets, and about 745 million word senses
  - Obtained by automatic integration of Wikipedia (encyclopedic) and WordNet (lexicographic)

## English WordNet Interface

### WordNet Search - 3.1

- WordNet home page - Glossary - Help

Word to search for: boy Search WordNet

Display Options: (Select option to change) ▼ Change

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

#### Noun

- <u>S:</u> (n) <u>male child</u>, **boy** (a youthful male person) "the baby was a boy"; "she made the boy brush his teeth every night"; "most soldiers are only boys in uniform"
- S: (n) boy (a friendly informal reference to a grown man) "he likes to play golf with the boys"
- <u>S:</u> (n) <u>son</u>, **boy** (a male human offspring) "their son became a famous judge"; "his boy is taller than he is"



### English WordNet Interface contd..

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- WordNet home page - Glossary - Help

Word to search for: boy Search WordNet

Display Options: (Select option to change) ▼ Change

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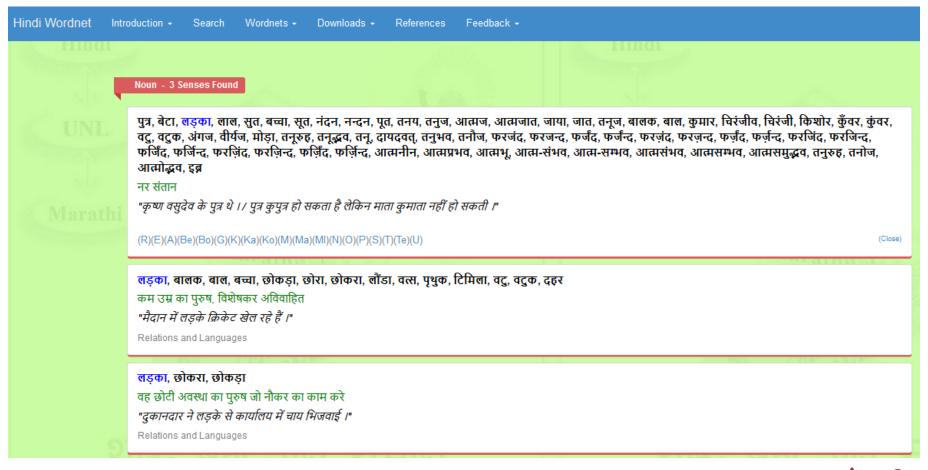
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#### Noun

- <u>S:</u> (n) <u>male child</u>, **boy** (a youthful male person) "the baby was a boy"; "she made the boy brush his teeth every night"; "most soldiers are only boys in uniform"
  - o <u>direct hyponym</u> / <u>full hyponym</u>
  - o <u>direct hypernym</u> I <u>inherited hypernym</u> I <u>sister term</u>
    - S: (n) male, male person (a person who belongs to the sex that cannot have babies)
  - antonym
    - <u>W:</u> (n) <u>female child</u> [Opposed to: <u>male child</u>] (a youthful female person) "the baby was a girl"; "the girls were just learning to ride a tricycle"
    - W: (n) girl [Opposed to: boy] (a youthful female person) "the baby was a girl"; "the girls were just learning to ride a tricycle"
  - o derivationally related form

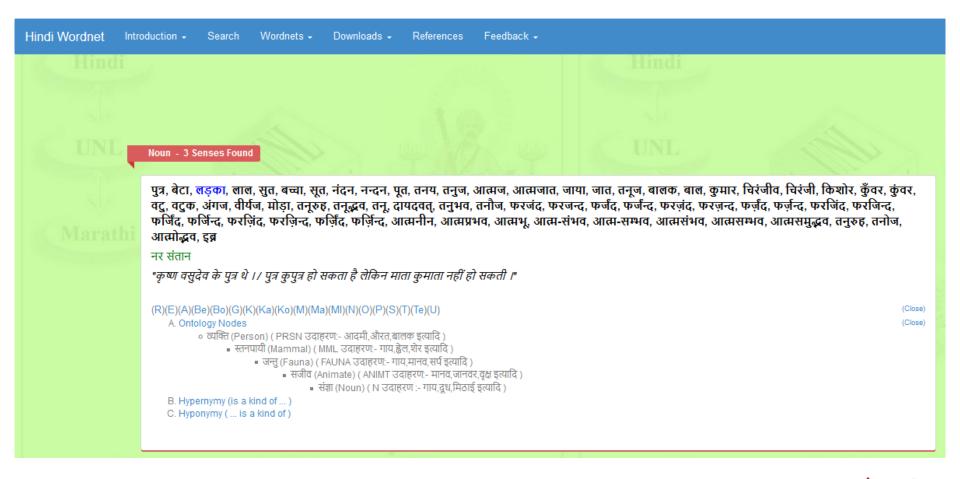


### Hindi WordNet Interface



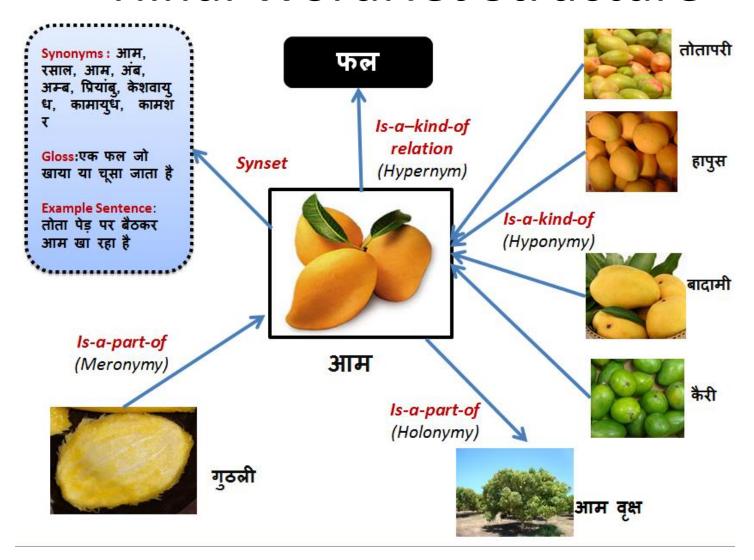


### Hindi WordNet Interface contd..



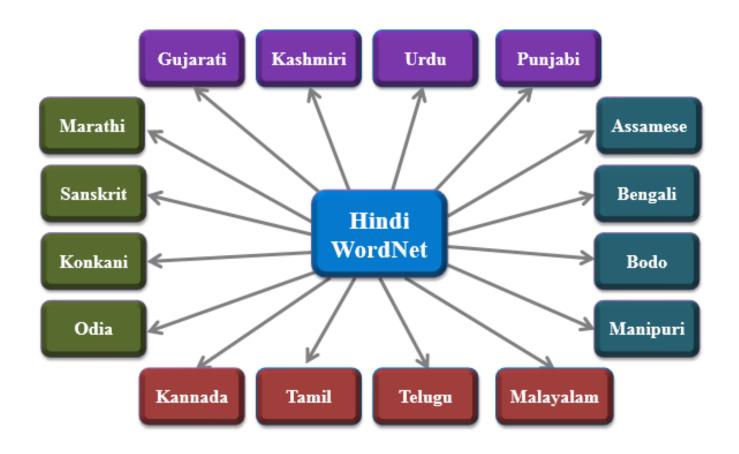


### Hindi WordNet Structure



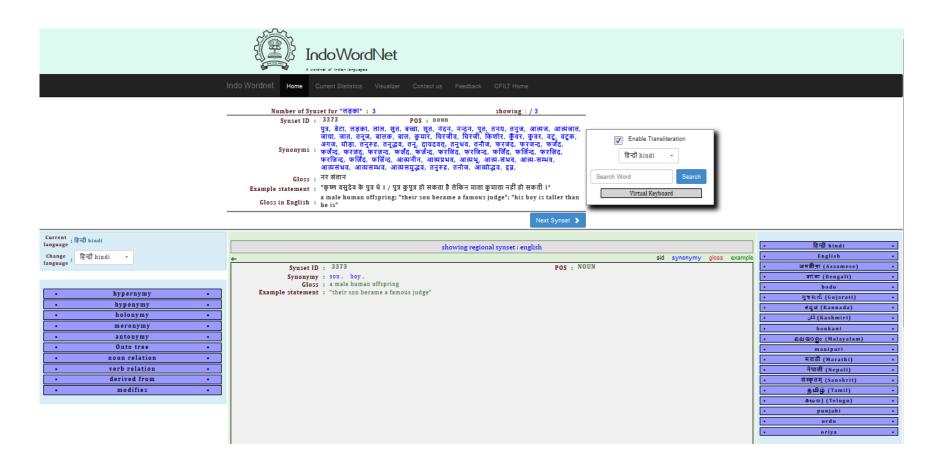


### IndoWordNet Structure





### IndoWordNet Interface





## IndoWordNet linked Synset

(4265) (n)

ছেলে, বালক

কম ব্য়সের পুরুষ, বিশেষত অবিবাহিত

"भग़पाल (ছलता क्रिकिट (थलएह" (4265) (n)

लड़का, बालक, बाल, बच्चा, छोकड़ा, छोरा, छोकरा

कम उम्र का पुरुष, विशेषकर अविवाहित

"मैदान में लड़के क्रिकेट खेल रहे हैं ।" (4265) (n)

मुलगा,पोरगा,पोर,पोरगे

साधारणतः सोळा वर्षांखालील पुरुष व्यक्ती

"तो मुलगा खूपच हुशार आहे"

Bengali WordNet Hindi WordNet Marathi WordNet



# IndoWordNet Synset Statistics

	Noun	Verb	Adjective	Adverb	Total
Hindi	29664	3626	6313	534	40137
Assamese	9065	1676	3805	412	14958
Bengali	27281	2804	5815	445	36346
Bodo	8788	2296	4287	414	15785
Gujarati	26503	2805	5828	445	35599
Kannada	12765	3119	5988	170	22042
Kashmiri	21041	2660	5365	400	29469
Konkani	23144	3000	5744	482	32370
Malayalam	20071	3311	6257	501	30140
Manipuri	10156	2021	3806	332	16351
Marathi	23271	3146	5269	539	32226
Nepali	6748	1477	3227	261	11713
Odiya	27216	2418	5273	377	35284
Punjabi	23255	2836	5830	443	32364
Sanskrit	31476	1247	4004	265	36997
Tamil	16312	2803	5827	477	25419
Telugu	12078	2795	5776	442	21091
Urdu	22990	2801	5786	443	34280



### IndoWordNet Visualizer Interface

### **IndoWordNet Visualizer**

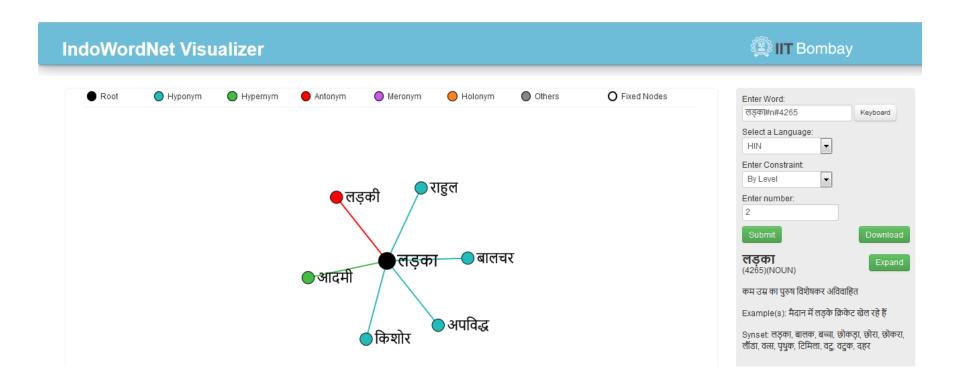


Sense ID	PoS	Meaning	Example	Synset
3373		नर संतान	"कृष्ण वसुदेव के पुत्र थे/ पुत्र कुपुत्र हो सकता है लेकिन माता कुमाता नहीं हो सकती"	पुत्र, बेटा, लड़का, लाल, सुत, बच्चा, सूत, नंदन, नन्दन, पूत, तनय, तनुज, आत्मज, आत्मजात, तनूज, बालक, कुमार, चिरंजीव, चिरंजी, किशोर, वटु, वटुक, अंगज, मोड़ा, तनूरुह, तनूद्भव, तनू, दायदवत्, तनुभव, तनौज, फरजंद, फरजिंद, आत्मनीन, आत्मप्रभव, आत्मभू, आत्म-संभव, आत्म-सम्भव, आत्मसंभव, आत्मसम्भव, आत्मसमुद्भव, तनुरुह, तनोज, आत्मोद्भव, इब्र
5896	NOUN	वह छोटी अवस्था का पुरुष जो नौकर का काम करे	"दुकानदार ने लड़के से कार्यालय में चाय भिजवाई"	लड़का, छोकड़ा, छोकरा
4265	NOUN	कम उम्र का पुरुष विशेषकर अविवाहित	"मैदान में लड़के क्रिकेट खेल रहे हैं"	लड़का, बालक, बच्चा, छोकड़ा, छोरा, छोकरा, लौंडा, वत्स, पृथुक, टिमिला, वटु, वटुक, दहर





### IndoWordNet Visualizer Interface



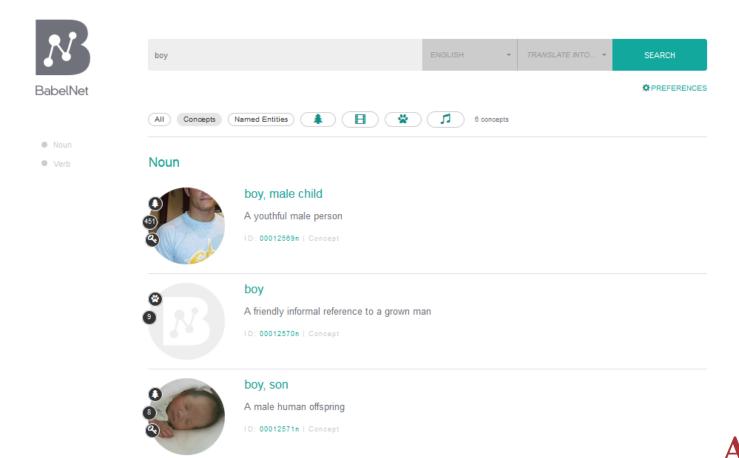


### IndoWordNet Visualizer Interface

#### **Bombay** IndoWordNet Visualizer Root Hyponym Hypernym Meronym Holonym Others Fixed Nodes Antonym Enter Word: लंडका#n#4265 Keyboard ● विवाहि**—** महापुरुष ● गड़दार 🔵 अपविद्ध Select a Language: 🔵 बालचर 🔵 भानजा 🔵 सुंदर\_पुरुष Enter Constraint: ्राहुल By Level **ा**फाहि**ग्रा**म्बौना **ा**ग्वाला किशोरी • मुमूर्षे ्ताऊ <u>\_</u>किशोर Enter number: 🔵 शांभवी भील ुभ्<sub>सुर</sub> विधुर • पोलो लड़का ्मुछड़ कायर पुरुषायक ঝিলাভিন <u>्र</u>कुमारी लड़का सिंहिका (4265)(NOUN) आजागुरु नालबंद **्**लंडकी अदिर्म **●**महिला अरक्षणी कम उम्र का पुरुष विशेषकर अविवाहित <u>वर</u> **ि**षिता <mark>●सँपेरा</mark> माली अन्यपूर्वा Example(s): मैदान में लड़के क्रिकेट खेल रहे हैं **्**घसियारा भतीजा हजाम 🔵 मेहतर 🌎 अधम\_पुरुष Synset: लंडका, बालक, बच्चा, छोकडा, छोरा, छोकरा, वाग्दत्ता लौंडा, वत्स, पृथक, टिमिला, वट, वटक, दहर अन्यपुरुष् लक् इहारा 👅 नागकन्य्रा 🔵 कुरूप\_पुरुष 🔵 अलका 🔵 उसमान 🍗 गृहस्थ\_आश्रमी



### **BabelNet**



SCHOOL OF DATA SCIENCE SCHOOL OF TELECOMMUNICATION

### Wordnets in the World

 The Global WordNet Organization gives access of wordnets in the world

http://globalwordnet.org/wordnets-in-the-world/

 Albanian, Arabic, Spanish, Catalan, Basque, Italian, Bulgarian, Czech, Greek, Romanian, Serbian, Turkish, Chinese, Danish, Dutch, Estonian, French, German, Hungarian, Icelandic, Portuguese, Irish, Japanese, Korean, Kurdish, Latin, Macedonian, Norwegian, Persian, Polish, Russian, Swedish



### WordNet Applications

- Machine Translation
- Word Sense Disambiguation
- Sentiment Analysis
- Information Retrieval
- MultiWord Expression Detection
- Document structuring and categorization
- Cognitive NLP



## NLP Python libraries

- NumPy (Mathematical Computing, Advanced mathematical functionalities)
- Matplotlib (Numerical plotting library, useful in data analysis)
- **Scipy** (Library for scientific computation)
- Scikit-learn (Machine Learning/Data-mining library,
- PIL (Python library for Image Processing)
- PySpeech (Library for speech processing and text-to speech conversion)
- XML/LXML (XML Parsing and Processing)
- NLTK (Natural Language Processing)
   And many more...



# The Natural Language Toolkit (NLTK)

- Developed by Steven Bird and Co. at Stanford University (2006).
- Open source python modules, datasets and tutorials
- Papers:
  - Bird, Steven. "NLTK: the natural language toolkit." *Proceedings of the COLING/ACL on Interactive presentation sessions*. Association for Computational Linguistics, 2006.
  - Loper, Edward, and Steven Bird. "NLTK: The natural language toolkit."
     *Proceedings of the ACL-02 Workshop on Effective tools and methodologies for teaching natural language processing and computational linguistics-Volume 1*. Association for Computational Linguistics, 2002.



### Components of NLTK

• Code: corpus readers, tokenizers, stemmers, taggers, chunkers, parsers, wordnet, ... (50k lines of code)

 Corpora: >30 annotated data sets widely used in natural language processing (>300Mb data)

Documentation: a 400-page book, articles, reviews,
 API documentation



# Components of NLTK

- Code
  - Corpus Readers
  - Tokenizers
  - Stemmers
  - Taggers
  - Parsers
  - WordNet
  - Semantic Interpretation
  - Clusterers
  - Evaluation Metrics



# Components of NLTK contd...

### Corpus

- Brown Corpus
- Carnegie Mellon Pronouncing Dictionary
- CoNLL 2000 Chunking Corpus
- Project Gutenberg Selections
- NIST 1999 Information Extraction: Entity Recognition Corpus
- US Presidential Inaugural Address Corpus
- Indian Language POS-Tagged Corpus
- Floresta Portuguese Treebank
- Prepositional Phrase Attachment Corpus
- SENSEVAL 2 Corpus
- Sinica Treebank Corpus Sample
- Universal Declaration of Human Rights Corpus
- Stopwords Corpus
- TIMIT Corpus Sample
- Treebank Corpus Sample



## Components of NLTK contd...

### Books:

- Natural Language Processing with Python Steven Bird, Edward Loper, Ewan Klein
- Python Text Processing with NLTK 2.0 Cookbook Jacob Perkins
- Included in NLTK:
  - Installation instructions
  - API Documentation: describes every module, interface, class, and method



### **NLTK Modules**

NLP Tasks	NLTK Modules	Functionality	
Accessing Corpora	nltk.corpus	Standardized interfaces to corpora and lexicons	
String Processing	nltk.tokenize, nltk.stem	Tokenizers, sentence tokenizers, stemmers	
Collection Discovery	nltk.collections	t-test, chi-squared, point-wise mutual information	
POS Tagging	nltk.tag	n-gram, backoff, Brill, HMM, TnT	
Chunking	nltk.chunk	Regular expression, n-gram, named entity	
Parsing	nltk.parse	Chart, feature-based, unification, probabilistic, dependency	
Classification	nltk.classify, nltk.cluster	Decision tree, maximum entropy, naive Bayes, EM, k-means	
Semantic Interpretation	nltk.sem, nltk.inference	Lambda calculus, first-order logic, model checking	
Evaluation Metrics	nltk.metrics	Precision, recall, agreement coefficients	
Probability Estimation	nltk.probability	Frequency distributions, smoothed probability distributions	
Applications	nltk.app	Graphical concordancer, parsers, WordNet browser	
Linguistics fieldwork	nltk.toolbox	Manipulate data in SIL Toolbox format	

## Accessing corpus using NLTK

```
>>> from nltk.corpus import brown
>>> brown.words()
['The', 'Fulton', 'County', 'Grand', 'Jury', 'said', ...
>>> from nltk.corpus import twitter_samples
>>> twitter_samples.fileids()
[u'negative_tweets.json', u'positive_tweets.json', u'tweets.20150430-223406.json']
```



### Accessing wordnet synsets

```
>>> from nltk.corpus import wordnet
>>> from nltk.corpus import wordnet as wn
>>> wn.synsets('dog')
[Synset('dog.n.01'), Synset('frump.n.01'), Synset('dog.n.03'),
```

[Synset('dog.n.01'), Synset('frump.n.01'), Synset('dog.n.03'), Synset('cad.n.01'), Synset('frank.n.02'), Synset('pawl.n.01'), Synset('andiron.n.01'), Synset('chase.v.01')]

```
>>> wn.synsets('dog', pos=wn.VERB)
[Synset('chase.v.01')]
```



### Accessing sense definition, lemma

```
>>> wn.synset('dog.n.01').definition()
```

a member of the genus Canis (probably descended from the common wolf) that has been domesticated by man since prehistoric times; occurs in many breeds

```
>>> wn.synset('dog.n.01').lemmas()
[Lemma('dog.n.01.dog'), Lemma('dog.n.01.domestic_dog'),
Lemma('dog.n.01.Canis_familiaris')]
```

```
>>> [str(lemma.name()) for lemma in wn.synset('dog.n.01').lemmas() ] ['dog', 'domestic_dog', 'Canis_familiaris']
```



### Accessing semantic relations

```
>>> dog = wn.synset('dog.n.01')

>>> dog.hypernyms()
[Synset('canine.n.02'), Synset('domestic_animal.n.01')]

>>> dog.hyponyms()
[Synset('basenji.n.01'), Synset('corgi.n.01'), Synset('cur.n.01'),
Synset('dalmatian.n.02'), ...]
```



### **WordNet Similarities**

```
>>> dog = wn.synset('dog.n.01')
>>> cat = wn.synset('cat.n.01')
>>> dog.path_similarity(cat)
0.2
>>> wn.lch_similarity(dog,cat)
2.0281482472922856
```

<sup>\*\*</sup>Leacock-Chodorow Similarity: Return a score denoting how similar two word senses are, based on the shortest path that connects the senses (as above) and the maximum depth of the taxonomy in which the senses occur"



<sup>\*\*</sup> Path similarity:Return a score denoting how similar two word senses are, based on the shortest path that connects the senses in the is-a (hypernym/hypnoym) taxonomy

### Access to the Open Multilingual WordNet

```
>>> sorted(wn.langs())
['als', 'arb', 'cat', 'cmn', 'dan', 'eng', 'eus', 'fas',
'fin', 'fra', 'fre', 'glg', 'heb', 'ind', 'ita', 'jpn', 'nno',
'nob', 'pol', 'por', 'spa', 'tha', 'zsm']
>>> wn.synsets(b'\xe7\x8a\xac'.decode('utf-8'), lang='jpn')
[Synset('dog.n.01'), Synset('spy.n.01')]
>>> wn.synset('dog.n.01').lemma_names('ita')
['cane', 'Canis_familiaris']
```

