



KALINGA INSTITUTE OF INDUSTRIAL TECHNOLOGY
Deemed to be University
BHUBANESWAR-751024

School of Computer Engineering
Spring Semester 2025

Natural Language Processing (NLP)-CS30016 (L-T-P-Cr:3-0-0-3)

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Course Objectives:

- To understand the steps involved in Natural language processing
- To learn about the lexical, syntactic and semantic analysis of natural language processing
- To explore the various parsing techniques for natural languages
- To understand the statistical models for Natural language processing
- To learn about the various applications involved in Natural language processing.

Day-Wise Lesson plan:-

| Lecture no. | Topics to be covered |
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| 1 | <ul style="list-style-type: none">• Introduction to NLP (With real-world examples) |
| 2 | <ul style="list-style-type: none">• Phases of NLP (Lexical and Morphological, Syntactic, Semantic, Discourse Integration, Pragmatic)• NLP Pipeline |
| 3 | <ul style="list-style-type: none">• NLP Applications• Latest NLP Tools• Demo with NLTK |

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| 4 | <ul style="list-style-type: none"> • Why is NLP hard? Ambiguity in NLP, Various types of ambiguity with examples • Empirical Laws (Zip's Law and Heap's Law) (Numerical and Conceptual Examples) |
| 5 | <ul style="list-style-type: none"> • What is text processing? Need for normalization and cleaning: Unicode, encoding, whitespace, punctuation, case-folding • Regular Expression and String Matching, Pattern matching |
| 6 | <ul style="list-style-type: none"> • Phonology and Morphology • Phonemes, syllables (phonology basics) • Morphs, morphemes, affixes • Word formation • Why does morphological understanding matter for stemming/ lemmatization? |
| 7 | <ul style="list-style-type: none"> • Tokenization • Sentence Segmentation • Word tokenization • Subword tokenization (Intro to BPE and WordPiece with simple numerical examples) • Stemming and Lemmatization • Stemming algorithms (Intro to Porter Stemmer) • Lemmatization • Stop Words Removal |
| 8 | <ul style="list-style-type: none"> • Part-of-Speech Tagging (POS: NN, VB, JJ, RB, etc.) • Intro to Rule-based, HMM-based, Neural POS taggers • Named Entity Recognition (NER) • Intro to Sequence labeling |
| 9 | <ul style="list-style-type: none"> • Language Modeling: Bi-gram, Tri-gram, N-gram Model (numerical examples of probabilistic models) • Intro to Noisy Channel Model and Smoothing (only basic Laplace and primary Backoff smoothing with numerical examples) |
| 10 | <ul style="list-style-type: none"> • Basics of Spelling Correction. (Numerical examples with Edit Distance-Levenshtein distance only) |
| 11 | <ul style="list-style-type: none"> • What is the word Feature Extraction and Representation? Word Embedding (Text to Vector) • BoW, CBoW |
| 12 | <ul style="list-style-type: none"> • TF-IDF (Conceptual and numerical examples) |

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| | <ul style="list-style-type: none"> Intro to Word2Vec with examples (CBoW and Skipgram intro with examples) |
| 13 | <ul style="list-style-type: none"> Hands-on Exercise on Text Pre-Processing (Using NLTK, Spacey) |
| 14 | <ul style="list-style-type: none"> Text Classification: Naive Bayes and Sentiment Classification Naive Bayes Classifier |
| 15 | <ul style="list-style-type: none"> Training the Naive Bayes Classifier |
| 16 | <ul style="list-style-type: none"> Hands-on Exercise on Naive Bayes Classifier |
| 17 | <ul style="list-style-type: none"> Other Text Classification tasks using Naive Bayes |
| 18 | <ul style="list-style-type: none"> Evaluation: Confusion Matrix, Precision, Recall, F-Measure |
| 19 | <ul style="list-style-type: none"> Hands-on Exercise on Evaluation of Text Classifier |
| 20 | <ul style="list-style-type: none"> Tutorials/ Activity/ Project/ Assignment: NLP Applications: Chatbot, Classifier, Spam Detector, Sentiment Analysis from X or Social Media, Projects based on NLP Pipeline |
| <i>MID SEMESTER</i> | |
| 21 | <ul style="list-style-type: none"> Markov Models, Hidden Markov Models (HMMs) |
| 22 | <ul style="list-style-type: none"> Viterbi Algorithm |
| 23 | <ul style="list-style-type: none"> Estimating the parameters of HMMs |
| 24 | <ul style="list-style-type: none"> The Forward-Backward Algorithm |
| 25 | <ul style="list-style-type: none"> Implementation Issues |
| 26 | <ul style="list-style-type: none"> HMM Tagging, Word Segmentation and NER |
| 27 | <ul style="list-style-type: none"> Intro to RNN (Learning meaning of words, not just probabilities) Advantages of RNN over HMM. Comparison |

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| 28 | <ul style="list-style-type: none"> ● RNN: Recurrent Hidden State (Dependency Learning) |
| 29 | <ul style="list-style-type: none"> ● RNN: Gating Mechanisms (LSTM/ GRU) for long term dependency learning |
| 30 | <ul style="list-style-type: none"> ● Attention Mechanism |
| 31 | <ul style="list-style-type: none"> ● Self Attention |
| 32 | <ul style="list-style-type: none"> ● Multi-head attention |
| 33 | <ul style="list-style-type: none"> ● Transformers Encoder (BERT) |
| 34 | <ul style="list-style-type: none"> ● Transformers Decoder (Cross Attention) |
| 35 | <ul style="list-style-type: none"> ● Encoder-Decoder Architecture |
| 36 | <ul style="list-style-type: none"> ● Intro to Speech and its role in NLP (Biology of Human Speech Production, From Waveform to Spectrogram) ● Speech vs Text in NLP |
| 37 | <ul style="list-style-type: none"> ● Speech Signal Processing Essentials (Framing and Windowing, Frequency Domain Basics, Speech Features) |
| 38 | <ul style="list-style-type: none"> ● Feature Engineering vs Deep Learning (MFCC vs wav2vec features) |
| 39 | <ul style="list-style-type: none"> ● Intro to Automatic Speech Recognition Pipeline with examples |
| 40 | <ul style="list-style-type: none"> ● Hands-on Exercise on Speech Processing |
| Tutorials / Activity/ Project/ Assignment: Transformer based models, RNN, LSTM, GRU, BERT, Speech Processing. | |

Course Outcome:

At the end of the course, the students will be able to:

CO1: Understand the concepts of NLP and its algorithms.

CO2: Evaluate different computing architectures for natural language processing for various parameters.

CO3: Apply the different modelling and tagging concepts for language processing.

CO4: Analyze the various grammars & parsing algorithms.

CO5: Explore the role of statistical parsing & machine translation.

CO6: Implementation of the role of natural language processing in real life applications.

Activity Calendar:

| Name | Tentative Date | Marks |
|---------------------------------------|-----------------|-------|
| Conceptual Assignment | 17 Jan 2026 | 5 |
| Class Test (Syllabus till Lecture 12) | 19/ 20 Jan 2026 | 5 |
| Programming Assignment 1 | 25 Jan 2026 | 5 |
| Programming Assignment 2 | 13 March 2026 | 5 |
| Class Notes Check | 23/ 24 Mar 2026 | 5 |
| Class Performance and Attendance | 27 Mar 2026 | 5 |

Text Books:

1. Christopher Manning, Schutze Heinrich, “Foundations of Statistical Natural Language Processing”, MIT Press, 1999.
2. Speech and Language Processing, Jurafsky, D. and J. H. Martin, Prentice-Hall.

Reference Books:

1. Natural Language Understanding, Allen, J., The Benjamins/Cummings Publishing Company Inc.
2. Statistical Methods for Speech Recognition, Jelinek, F., The MIT Press.

Online Resources:

1. NPTEL Course by Prof. Pawan Goyal:
<https://nptel.ac.in/courses/106105158>
2. Community curated list of NLP Resources:
<https://github.com/keon/awesome-nlp>
3. High Quality Resources for NLP:
<https://medium.com/nlplanet/awesome-nlp-18-high-quality-resources-for-studying-nlp-1b4f7fd87322>
4. Stanford NLP Course:
<https://web.stanford.edu/class/cs224n/>