3. Bernd Jahne, **Digital Image Processing**, (5th revised and extended edition), Springer, 2002

4. S. Annadurai and R. Shanmugalakshmi, **Fundamentals of Digital Image Processing**, Pearson Education, 2007

5. M.A. Joshi, **Digital Image Processing: An Algorithmic Approach**, Prentice-Hall of India, 2006

6. B. Chanda and D.D. Majumder, **Digital Image Processing and Analysis**, Prentice-Hall of India, 2007

MCSE202: COMPILER DESIGN

Course Objectives: The course aims to develop the ability to design, develop, and test a functional compiler/ interpreter for a subset of a popular programming language.

Course Learning Outcomes:

On completing this course, the student will be able to:

CO1: describe how different phases of a compiler work.

CO2: implement top-down and bottom-up parsing algorithms.

CO3: use tools like Lex and Yacc to implement syntax-directed translation.

Syllabus:

Unit- I Lexical and Syntactic Analysis: Review of regular languages, design of a lexical analyzer generator, context-free grammars, syntactic analysis: top-down parsing: recursive descent and predictive parsing, LL(k) parsing; bottom-up parsing: LR parsing, handling ambiguous in bottom-up parsers.

Unit-II Syntax directed translation: Top-down and bottom-up approaches, data types, mixed mode expression; subscripted variables, sequencing statement, subroutines and functions: parameters calling, subroutines with side effects.

Unit-III Code generation, machine dependent and machine-independent optimization techniques.

Readings:

1. Alfred V. Aho, Ravi Sethi, D. Jeffrey Ullman, Monica S. Lam, **Principles, Techniques and Tools**, Pearson Education India, 2nd edition,, 2013.
2. A.V. Aho, M. S. Lam, R. Sethi and J. D. Ullman, **Compilers,** Pearson, 2016.
3. Dick Grune, Kees van Reeuwijk, Henri E .Bal, Ceriel J.H. Jacobs, K Langendoen, **Modern Compiler Design,** Springer, 2012.

MCSE 203: NATURAL LANGUAGE PROCESSING [3-0-1]

Course Objectives: The course provides a rigorous introduction to the essential components of a Natural Language Processing (NLP) system. The students will learn various statistical, machine learning, and deep learning techniques in NLP and apply them to solve machine translation and conversation problems.

Course Learning Outcomes:

On completing this course, the student will be able to:

CO1:Understand and describe the evaluation of NLP systems.

CO2:Understand deep learning techniques in NLP and apply them to solve machine translation and conversation problems.

CO3:Learn about major NLP issues and identify possible future areas of NLP research.

Syllabus:

UNIT I Introduction: Natural Language Processing (NLP), History of NLP, Neural Networks for NLP, Applications - Sentiment Analysis, Spam Detection, Resume Mining, Conversation Modeling, Chat-bots, dialog agents, Question Processing

UNIT II Language Modeling and Part of Speech Tagging: Unigram Language Model, Bigram, Trigram, N-gram, Advanced smoothing for language modeling, Empirical Comparison of Smoothing Techniques, Applications of Language Modeling, Natural Language Generation, Parts of Speech Tagging, Morphology, Named Entity Recognition

UNIT III Words and Word Forms: Bag of words, skip-gram, Continuous Bag-Of-Words, Embedding representations for words Lexical Semantics, Word Sense Disambiguation, Knowledge Based and Supervised Word Sense Disambiguation

UNIT IV Text Analysis, Summarization and Extraction: Sentiment Mining, Text Classification, Text Summarization, Information Extraction, Named Entity Recognition, Relation Extraction, Question Answering in Multilingual Setting; NLP in Information Retrieval, Cross-Lingual IR

UNIT V Machine Translation: Need of MT, Problems of Machine Translation, MT Approaches, Direct Machine Translations, Rule-Based Machine Translation, Knowledge Based MT System, Statistical Machine Translation (SMT), Parameter learning in SMT (IBM models) using EM), Encoder-decoder architecture, Neural Machine Translation

Readings:

1. Speech and Language Processing. Dan Jurafsky and James H. Martin. Pearson (2009).
2. Introduction to Natural Language Processing. Jacob Eisenstein. The MIT Press (2019).
3. Neural Network Methods for Natural Language Processing. Yoav Goldberg. Morgan and Claypool Publisher (2017).
4. Deep Learning for Natural Language Processing: Develop Deep Learning Models for Natural Language in Python. Jason Brownlee. Machine Learning Mastery (2019).
5. Natural Language Processing with Python: Analyzing Text with the Natural Language Toolkit. Steven Bird, Ewan Klein and Edward Loper. O'Reilly (2009).