# **Cmpe 493 Introduction to Information Retrieval**

There has been a striking growth in text data such as web pages, news articles, e-mail messages, social media data, and scientific publications in the recent years. Developing tools for accessing, managing, and utilizing this huge amount of textual information is getting increasingly important. This course will cover the technology underlying search engines, focusing on a wide range of topics including methods for processing, indexing, querying, and organizing textual data, as well as methods for web search, crawling, and link analysis.

Instructor: Arzucan Özgür

# **Course Objectives:**

- Understand how search engines work
- Learn to process, index, retrieve, and analyze textual data
- Learn to evaluate information retrieval systems
- Learn about web search, crawling and link analysis
- Build working systems that help users find useful information on the Web
- Learn about the state of the art in information retrieval research

**Web site:** Course content will be available at Moodle (http://moodle.cmpe.boun.edu.tr/). You should subscribe with your "boun" email address and the key provided in the first lecture

#### **Textbook:**

Christopher D. Manning, Prabhakar Raghavan and Hinrich Schütze, *Introduction to Information Retrieval*, Cambridge University Press. 2008. http://nlp.stanford.edu/IR-book/information-retrieval-book.html

### **Reference books (Optional):**

Daniel Jurafsky and James H. Martin, SPEECH and LANGUAGE PROCESSING: An Introduction to Natural Language Processing, Computational Linguistics, and Speech Recognition, Second Edition, Prentice-Hall, 2008.

Christopher D. Manning and Hinrich Schütze, *Foundations of Statistical Natural Language Processing*, MIT Press, 1999. <a href="http://nlp.stanford.edu/fsnlp/">http://nlp.stanford.edu/fsnlp/</a>

# **Tentative List of Topics:**

- Boolean model; text pre-processing; inverted indexes
- Approximate string matching and tolerant retrieval
- Index construction and compression
- Vector space model; text-similarity metrics; term weighting; ranked retrieval
- Evaluating information retrieval systems
- Relevance feedback; query expansion
- Language models for information retrieval
- Text classification and clustering

- Latent semantic indexing
- Web search and crawling
- Link analysis (e.g. hubs and authorities, Google PageRank)

# **Course Requirements:**

The lectures will take place on Wednesdays between 11:00-12:00 and Thursdays between 15:00-17:00. You are encouraged to attend and actively participate in the lectures.

The homework assignments will involve some programming where you will implement and test some of the techniques that we cover in class. You can use any programming language of your choice such as Java, Python, Perl, and etc.

The midterm and final exams will consist of problems covering the material in the lectures.

Each student (individually or in pairs) will be responsible for designing and completing a research project that demonstrates the ability to use concepts from the class. Each team will choose a project topic by selecting a recent scientific paper from an Information Retrieval conference or journal. The project will involve addressing the limitations of the existing work, and extending or replicating it in some way. The teams will give short project presentations in front of the class and submit a project report written in the format of a scientific paper in the end of the semester.

# **Grading:**

Midterm Exam: 15%Final Exam: 15%

- 2-3 Homework Assignments: 30%

- Project: 35%

- Class Participation: 5%