

### Department of Computer Science

## CS-4051 Information Retrieval Spring 2024

**Instructor Name:** Dr. Asma Naseer **TA Name:** 

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Office Location/Number: New Building - 11
Office Hours: Thu 10 pm - 12 pm

**Course Information** 

Program: BS (CS) Credit Hours: 3 Course Type: Elective

Pre-requisites (if any): Core Programming and Algorithm skills, Probability and Statistics

**Class Meeting Time:** (Mon & Wed) (8:30 am to 10:00 AM – 10:00 am 11:30 am)

Class Venue: CS-5

### **Course Description/Objectives/Goals:**

This course provides broad coverage of the important issues in information retrieval. It is designed to help you to understand how search engines work, how to build your own search engine, evaluate its performance, and modify it for specific applications. A number of advanced topics will be covered to address more recent developments in IR such as collaborative filtering and Topic Modeling. Students will furthermore acquire practical experience in the construction of IR systems by a series of programming assignments. Mathematical experience including basic probability is strongly desirable.

### **Course Learning Outcomes (CLOs):**

Domain	BT* Level
С	2
С	6
С	3
С	3
С	3
A	5
P	6
	C C C A

<sup>\*</sup> BT= Bloom's Taxonomy, C=Cognitive domain, P=Psychomotor domain, A= Affective domain

**Bloom's taxonomy Levels:** 1. Knowledge, 2. Comprehension, 3. Application, 4. Analysis, 5. Synthesis, 6. Evaluation

#### **Course Textbook**

[MRS] Introduction to Information Retrieval by Manning, Raghavan, and Schütze - available free online.

# Additional references and books related to the course:

- o (MG) Managing Gigabytes, by I. Witten, A. Moffat, and T. Bell.
- o (IRAH) Information Retrieval: Algorithms and Heuristics, by D. Grossman and O. Frieder.
- o (MIR) Modern Information Retrieval, by R. Baeza-Yates and B. Ribeiro-Neto.
- o (FSNLP) Foundations of Statistical Natural Language Processing, by C. Manning and H. Schütze.
- o (SE) Search Engines: Information Retrieval in Practice, by B. Croft, D. Metzler, and T. Strohman.
- o (IRIE) Information Retrieval: Implementing and Evaluating Search Engines, by S. Büttcher, C. Clarke, and G. Cormack.
- o (TDMA) Text Data Management and Analysis: A Practical Introduction to Information Retrieval and Text Mining, C. Zhai and S. Massung, ACM Book Series, Morgan & Claypool Publishers, 2016.

### **Tentative Weekly Schedule**

Week	Topics	Text Book Sections
1	Key problems, Information need, Queries and documents, Matching scores	[MRS] Chapter 1
2	Text Preprocessing Tokenization Stopping, stemming	[MRS] Chapter 2 Section 2.2
3	Inverted Index Construction (Posting Lists, Dictionary, Distributed indexing, dynamic indexing)  • Term frequency (TF)	[MRS] Chapter 4 (from 4.2 till 4.5)
	<ul> <li>Document frequency (DF) and inverse document frequency (IDF)</li> </ul>	
	TF transformation	
	• BM25	
	Inverted index and postings	
	• Binary coding, unary coding, gamma-coding, and d-gap <i>Recommended Readings:</i>	
	<ul> <li>⇒ C. Zhai and S. Massung, Text Data Management and Analysis: A Practical Introduction to Information Retrieval and Text Mining, ACM Book Series, Morgan &amp; Claypool Publishers, 2016. Chapter 6 - Section 6.3, and Chapter 8.</li> </ul>	
	⇒ Ian H. Witten, Alistair Moffat, and Timothy C. Bell. Managing Gigabytes: Compressing and Indexing Documents and Images, Second Edition. Morgan Kaufmann, 1999.	
4	Zipf's Law, Heap's Law, Index Compression	[MRS] Chapter 5 Section 5.1 and 5.3
5	Retrieval Models (Vector Space Models) (Vector-space model, Cosine Similarity, Tf-Idf, BM25)  • Query likelihood	[MRS] Chapter 6 Section 6.2 and 6.3
	Statistical and unigram language models	
	Maximum likelihood estimate	
	Background, collection, and document language models	
	Smoothing of unigram language models	

	Relation between query likelihood and TF-IDF weighting	
	<ul> <li>Linear interpolation smoothing</li> </ul>	
	Feedback in the Vector Space Model	
	Feedback in Language Model	
	Recommended Readings:	
	C. Zhai and S. Massung, Text Data Management and	
	Analysis: A Practical Introduction to Information Retrieval and Text Mining, ACM Book Series, Morgan & Claypool Publishers, 2016. <b>Chapter 6 - Section 6.4</b>	
6	Word Vectors (Word Embeddings)	
7	Retrieval Models ( Language Models) Smoothing Methods	[MRS] Chapter 12 12.1 to 12.3
8	IR Evaluation/ Measures (Ranking measures: R-prec, Mean Average Precision, nDCG, Reciprocal Rank)  • Evaluation methodology	[MRS] Chapter 8 Section 8.1 to 8.4 (Interpolated precision is not included)
	Precision and recall	
	<ul> <li>Average precision, mean average precision (MAP), and geometric mean average precision (gMAP)</li> </ul>	
	Reciprocal rank and mean reciprocal rank	
	• F-measure	
	Normalized Discounted Cumulative Gain (nDCG)	
	Statistical significance test	
	Recommended Readings:	
	⇒ Mark Sanderson. Test collection based evaluation of information retrieval systems. Foundations and Trends in Information Retrieval 4, 4 (2010), 247-375.	
	⇒ C. Zhai and S. Massung, Text Data Management and Analysis: A Practical Introduction to Information Retrieval and Text Mining, ACM Book Series, Morgan & Claypool Publishers, 2016. Chapter 9	
9	Web Retrieval (Link analysis, Markov Chains, PageRank)  • Relevance feedback	[MRS] Chapter 21 21.1, 21.2.1, 21.2.2 [CMS]4.5.2
	Pseudo-relevance feedback	
	Implicit feedback	
	Rocchio feedback	
	Scalability and efficiency	
	• Spams	
	Crawler, focused crawling, and incremental crawling	
	• Google File System (GFS)	
	<ul> <li>MapReduce</li> </ul>	
	Link analysis and anchor text	

	PageRank and HITS	
	Recommended Readings:	
	⇒ C. Zhai and S. Massung, Text Data	
	Management and Analysis: A Practical Introduction to Information Retrieval and Text Mining, ACM Book Series, Morgan & Claypool Publishers, 2016. <b>Chapters 7 &amp; 10</b>	
10	Logistic Regression  • Learning to rank, features, and logistic regression	
	Content-based filtering	
	Collaborative filtering	
	Beta-Gamma threshold learning	
	User profile	
	Exploration-exploitation tradeoff	
11	Document Clustering (K-means clustering, Evaluation of clustering)  Overview of Clustering Techniques	[MRS] Chapter 16 Section 16.3, 16.4
	Document Clustering	
	Term Clustering	
	Evaluation of Text Clustering	
12	Clustering (Hierarchal Agglomerative Clustering )	[MRS] Chapter 17 Section 17.1 till 17.4
13	Text Classification (Naive Bayes, Sentiment Analysis)  • Overview of Text Categorization/Classification Methods	[MRS] Chapter 13 Section 13.1 till 13.3
	Text Categorization Problem	
	Features for Text Categorization	
	Classification Algorithms	
	Evaluation of Text Categorization	
14	Project Presentations	

### (Tentative) Grading Criteria:

- 1. Assignments + Class Exercises (10%)
- 2. Quizzes (10%)
- 3. Project and research paper (10+5%)
- 4. Midterm Exam (25%)
- 5. Final Exam (40%)
- Grading scheme for this course is **Absolute** under application of CS department's grading policies.
- o Minimum requirement to pass this course is to obtain at least 50% absolute marks

### **Course Policies:**

- o All assignments and homework must be done individually, until specified as a group task.
- o Quizzes will be announced.
- o No makeup for missed quiz or assignment.
- o Late Submissions of assignments will not be accepted.
- o Minimum 80% attendance is required for appearing in the Final exams.

# **Plagiarism in Assignments**

You are not allowed to copy code for programming assignments from internet or any other student. Penalty of plagiarism in programming assignments will be from one of the following depending on severity of case:

- o -1 absolute from final grade
- o Final grade is lowered
- o F in course