

Wright State University

CORE Scholar

Computer Science & Engineering Syllabi

College of Engineering & Computer Science

Winter 2010

CS 771-01: Natural Language Processing Techniques

Shaojun Wang

Wright State University - Main Campus, shaojun.wang@wright.edu

Follow this and additional works at: https://corescholar.libraries.wright.edu/cecs_syllabi



Part of the [Computer Engineering Commons](#), and the [Computer Sciences Commons](#)

Repository Citation

Wang, S. (2010). CS 771-01: Natural Language Processing Techniques. .
https://corescholar.libraries.wright.edu/cecs_syllabi/792

This Syllabus is brought to you for free and open access by the College of Engineering & Computer Science at CORE Scholar. It has been accepted for inclusion in Computer Science & Engineering Syllabi by an authorized administrator of CORE Scholar. For more information, please contact library-corescholar@wright.edu.

CS771: NATURAL LANGUAGE PROCESSING TECHNIQUES WINTER 2010

INFORMATION SYLLABUS ASSIGNMENTS PROJECT

TENTATIVE SYLLABUS

Day	Topic	Reading	Optional Reading
1/04/10	Introduction, basics of information theory	JM: 1; MS: 1-3	<u>Statistical methods</u> in Encyclopedia of Cognitive Science by Steve Abney.
1/06/10	Source-channel model, zipf's law and n-grams	JM: 9.1, 25.3, 4.4, 4.10-11; MS: 2.2.4, 1.4.3	<u>Information retrieval as statistical translation</u> Adam Berger and John Lafferty, SIGIR 1999
1/11/10	Smoothing techniques: Laplace, Good-Turing	JM: 4.5; MS: 6.2	Good, Jelinek, Mercer, and Robins on Turing's estimate of probabilities in American Journal of Mathematical and Management Sciences, 11, 229-308, 1991 by Arthur Nadas; <u>Always Good Turing: asymptotically optimal probability estimation</u> in Science. 302(5644):427-431 by Alon Orlitsky et al.
1/13/10	Deleted linear interpolation; EM algorithm for linear interpolation	JM: 4.6	<u>Interpolated Estimation of Markov Source Parameters from Sparse Data in Pattern Recognition in Practice</u> , 381-397, 1981 by F. Jelinek and R. Mercer; <u>Convexity, maximum likelihood and all that</u> By Adam Berger
1/18/10	Martin Luther King Day, University closed		
1/20/10	Back-off; Katz, Kneser-Ney; Limitations of n-grams	JM: 4.7-9; MS: 6.3	<u>A hierarchical Bayesian language model based on Pitman-Yor processes</u> by Yee Whye Teh, ACL 2006; <u>An empirical study of smoothing techniques for language modeling</u> in Computer Speech & Language, 13(4):319-358, 1999 by S. Chen and J.

			Goodman; <u>Up from trigrams! - the struggle for improved language models</u> . EUROSPEECH, 1037-1040, 1991 by Fred Jelinek
1/25/10	EM algorithm; Hidden Markov models	JM: 6; MS: 9	
1/27/10	Part-of-speech tagging, Named Entity recognition; Conditional random fields	JM: 6.22	<u>Conditional random fields: probabilistic models for sand labeling sequence data</u> by Lafferty et al., ICML 2001; <u>Identifying gene and protein mentions in text using conditional random fields</u> in Bioinformatics, 2005 by McDonald and Pereira
2/01/10	Semi-supervised conditional random fields	JM: 6	<u>Semi-supervised conditional random fields for improved sequence segmentation and labeling</u> by Jiao et al., ACL 2006; <u>Efficient computation of entropy gradient for semi-supervised conditional random fields</u> in NAACL/HLT, 2007 by Mann and McCallum
2/03/10	Grammars and parsing	JM: 12-13	
2/08/10	Probabilistic context- free grammars: inside-outside algorithm	JM: 14.1- 5; MS: 11	Lafferty's notes
2/10/10	Probabilistic lexicalized CFGs	JM: 14.6	<u>Head-driven statistics methods for natural language parsing</u> in Computational Linguistics, 29(4): 589-637, 2003 by Collins; <u>Intricacies of Collins' parsing model</u> in Computational Linguistics, 30(4): 479-511, 2004 by Bikel
2/15/10	Parser based language modeling: structured language model, generalized inside- outside algorithm		<u>Structured language model</u> in Computer Speech and Language, 14(4):283-332, 2000 by Chelba and Jelinek; <u>Stochastic analysis of structured language modeling</u> in Mathematical Foundations of Speech and Language Processing, 37-72, 2004 by Jelinek
2/17/10	Word sense disambiguation	JM: 20.1-5	<u>Unsupervised word sense disambiguation rivaling supervised methods</u> in ACL 1995 by Yarowsky
2/22/10	Topic models: LSA;	MS: 15.4	<u>Indexing by latent semantic</u>

	Topic based language modeling		analysis in Journal of the American Society for Information Science, 41(6):391-407, 1990 by Deerwester et al.; Exploiting latent semantic information in statistical language modeling in Proceedings of the IEEE, 88 (8):1279-1296, 2000 by Bellegarda
2/24/10	Topic models: PLSA; Composite language model		Unsupervised learning by probabilistic latent semantic analysis in Machine Learning, 42 (1):177-196, 2001 by Hoffman
3/01/10	Topic models: LDA		Latent Dirichlet allocation in Journal of Machine Learning Research, 3:993-1022, 2003 by Blei et al.
3/03/10	Machine translation: IBM models 1-5; phrase- and parsing based models	JM: 25.5, 25.11	The mathematics of statistical machine translation: parameter estimation in Computational Linguistics, 29(2): 263-311, 1993 by Brown et al.; Hierarchical phrase-based translation in Computational Linguistics, 33 (2):201-228, 2007 by Chiang; Statistical machine translation at Wikipedia
3/08/10	Machine translation: decoding algorithms	JM 25.8	
3/10/10	Machine translation: BLEU	JM 25.9	BLEU: a Method for Automatic Evaluation of Machine Translation in ACL 2002 by Papineni et al.; BLEU at Wikipedia
3/15/10	Project presentations		