Topics to cover (Reading list)

Reading list for the class (to appear in future reading assignments)

Improving the Q-matrix (skill to item mapping) from data:

Koedinger, K. R., McLaughlin, E. A., & Stamper, J. C. (2012). <u>Automated Student Model Improvement</u> (http://files.eric.ed.gov/fulltext/ED537201.pdf). International Educational Data Mining Society.

Lindsey, R. V., Khajah, M., & Mozer, M. C. (2014). <u>Automatic discovery of cognitive skills to improve the prediction of student learning (http://papers.nips.cc/paper/5554-automatic-discovery-of-cognitive-skills-to-improve-the-prediction-of-student-learning.pdf)</u>. In *Advances in neural information processing systems* (pp. 1386-1394).

Desmarais, M. (2010). <u>Conditions for effectively deriving a q-matrix from data with non-negative matrix factorization</u> (http://educationaldatamining.org/EDM2011/wp-content/uploads/proc/edm2011_paper35_full_Desmarais.pdf). In Educational Data Mining 2011.

Representation learning:

linear:

Adrian Colyer (2016) The Amazing Power of Word Vectors (blog (https://blog.acolyer.org/2016/04/21/the-amazing-power-of-word-vectors/)

Mikolov, T., Sutskever, I., Chen, K., Corrado, G. S., & Dean, J. (2013). <u>Distributed representations of words and phrases and their compositionality</u> (http://papers.nips.cc/paper/5021-distributed-representations-of-words-and-phrases-and-their-compositionality). In *Advances in neural information processing systems* (pp. 3111-3119).

Mikolov, T., Yih, W. T., & Zweig, G. (2013, June). <u>Linguistic regularities in continuous space word representations</u> (http://www.aclweb.org/anthology/N13-1#page=784). In hlt-Naacl (Vol. 13, pp. 746-751). Chicago

Pardos, Z.A., Dadu, A. (2017) Imputing KCs with Representations of Problem Content and Context (http://dl.acm.org/citation.cfm?id=3079689) [alt link (no paywall (http://people.ischool.berkeley.edu/~zp/papers/UMAP_kc_prediction.pdf)]. In Proceedings of the 25th Conferenceon User Modeling, Adaptation and Personalization (UMAP'17). Bratislava, Slovakia. ACM. Pp. 148-155.

Bolukbasi, T., Chang, K. W., Zou, J. Y., Saligrama, V., & Kalai, A. T. (2016). <u>Man is to computer programmer as woman is to homemaker? Debiasing word embeddings</u>
(http://papers.nips.cc/paper/6228-man-is-to-computer-programmer-as-woman-is-to-homemaker-debiasing-word-embeddings.pdf). In *Advances in Neural Information Processing Systems*(pp. 4349-4357).

non-linear:

Piech, C., Bassen, J., Huang, J., Ganguli, S., Sahami, M., Guibas, L. J., & Sohl-Dickstein, J. (2015). <u>Deep knowledge tracing (http://papers.nips.cc/paper/5654-deep-knowledge-tracing.pdf)</u>. In *Advances in Neural Information Processing Systems* (pp. 505-513).

Response: Khajah, M., Lindsey, R. V., & Mozer, M. C. (2016). <u>How deep is knowledge tracing?</u> (https://arxiv.org/pdf/1604.02416.pdf). arXiv preprint arXiv:1604.02416.

Pardos, Z.A., Tang, S., Davis, D., Le. C.V. (2017) <u>Enabling Real-Time Adaptivity in MOOCs with a Personalized Next-Step Recommendation Framework</u> (http://dl.acm.org/citation.cfm?id=3051471). In *Proceedings of the Fourth Conference on Learning @ Scale (L@S '17*). ACM. Pages 23-32.

Wang, L., Sy, A., Liu, L., Piech, C. (2017) <u>Learning to Represent Student Knowledge on Programming Exercises Using Deep Learning</u>

(http://educationaldatamining.org/EDM2017/proc_files/papers/paper_129.pdf). In *Proceedings of the 10th International Conference on Educational Data Mining*. Pp 324-329. Wuhan, China.

Ganin, Y., & Lempitsky, V. (2015). <u>Unsupervised domain adaptation by backpropagation</u> (http://sites.skoltech.ru/sites/compvision_wiki/static_pages//projects/grl/). In *International Conference on Machine Learning* (pp. 1180-1189).

Maaten, L. V. D., & Hinton, G. (2008). <u>Visualizing data using t-SNE</u> (http://www.jmlr.org/papers/volume9/vandermaaten08a/vandermaaten08a.pdf). Journal of Machine Learning Research, 9(Nov), 2579-2605.

Hinton, G. E., & Salakhutdinov, R. R. (2006). <u>Reducing the dimensionality of data with neural networks</u> (https://pdfs.semanticscholar.org/7d76/b71b700846901ac4ac119403aa737a285e36.pdf) . science, 313(5786), 504-507.