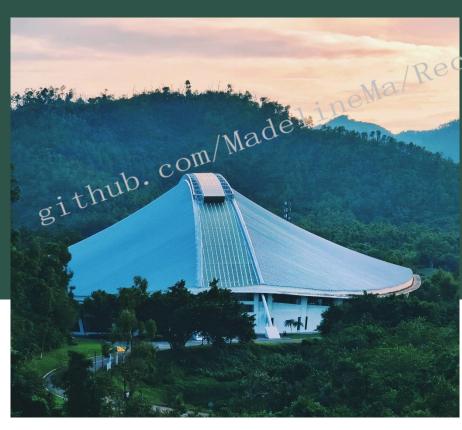




CONTENT

- 课程背景简介
- 课程内容与考核



Ma/Recommender-System

01 课程背景简介

- 课程名称:探索性大数据分析之推荐系统
- 推荐系统了解程度?
- 推荐系统直观感受?
- 推荐系统的增长引擎功能?

2019天猫双11 GMV 2684亿元,1%的转化率提升带来26.84亿元。 2018手淘惊喜红包,一毛撬动1.6元。

2019手淘618, PUSH带来免费流量1千万.

https://mp.weixin.qq.com/s/A36Oa81eku0vIX16OIMLOA https://mp.weixin.qq.com/s/q3kSWp5DTgo6i6vp3p9MuQ https://mp.weixin.qq.com/s/2oqCYiQGemCtSAkPoc5J6q



课程优势

- mender-System • 重点区别1: 1992年的传统且基础算法介绍与2015年起始的深度推荐系统研究;
- 重点区别2: 结合阿里巴巴淘系工作经验进行重点内容讲解;
- 深度推荐系统的必要性: 就业面试与科研;
- 自选难度与掌握程度: 传统推荐系统框架, 深度推荐系统框架, 传统算法, 深度学习算法 编程实现: ...儿化),表优劣之分;
 delineMa/Recomme
- 考核与自选难度保持一致: Survey, 算法的理解+代码实现+

课程难点

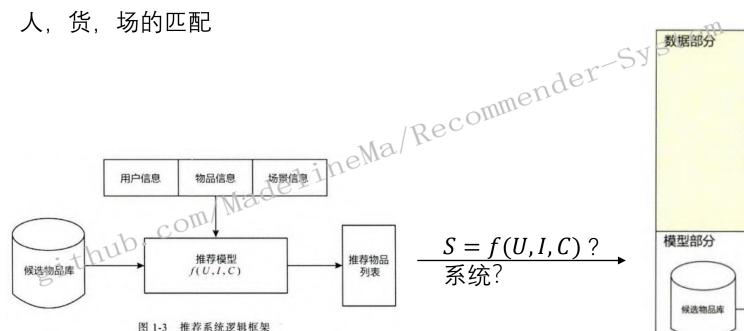
• 随堂给出补充机器学习知识的帖子+课后编程练习 需要多一些精力才能收获更多!

github.c





人,货,场的匹配



客户端及 流处理平台 大数据平台 服务器端 准实时数据 离线数据 实时数据 处理 处理 处理 用户信息 物品信息 场景信息 特征工程 用户特征 物品特征 场景特征 模型部分 Usystem 推荐系统模型 补充策略与 算法层 多样性 →推荐物品列表 候选物品库 实时性 召回层 ·流行度 ·新鲜度 ·冷启动 github.com/MadelineMag 线上A/B测试 模型在线更新 离线评估

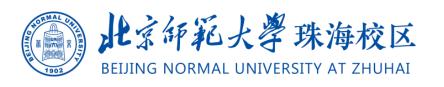
图 1-4 推荐系统的技术架构示意图

- Lesson 2: 半知 inema/Recommender—System
 Lesson 2: 半知 inema/Recommender—System • Lesson 2: 前深度学习时代 30%
- Lesson 3: 前深度学习时代 60%
 - Lesson 4: 前深度学习时代 100%
 - Lesson 5: 一个深度学习推荐系统的例子
 - Lesson 6: 深度机器学习知识补充

- Lesson 7: Python入门750%
 Lesson 8: Python入门 100%
 Lesson 9: Tensorfi-
 - Lesson 9: Tensorflow入门
 Lesson 10: Tensorflow实践
 - Lesson 11: Pytorch入门
 - Lesson 12: Pytorch实践

- Lesson 13: 多角度审视推荐系统 30% commender—System
 Lesson 14: 多角度审视推荐系统 30% commender—System
 Lesson 14: 多角度审视推荐系统 30% commender—System
- Lesson 15: 多角度审视推荐系统 100%
- Lesson 16: 推荐系统评估 50%
- · Lesson 17: 推荐系统评估 100%
- Lesson 18: 深度学习在推荐系统中的应用 AutoRec & Deep Crossing
- Lesson 19: 深度学习在推荐系统中的应用 NeuralCF & PNN
- Lesson 20: 深度学习在推荐系统中的应用 WDL & FM
- Lesson 21: 强化学习简介
- Lesson 22: 强化学习在推荐系统中的应用
- Lesson 23-28: 深度学习系统在推荐系统中的应用 6组(学生) Lesson 29-32: 前沿介绍
- Lesson 29-32: 前沿介绍

MadelineMa/Recommender-System



考核

• 平时成绩40%

reMa/Recommender-System 小组形式: 6小组进行22-27节的报告, 每组介绍30分钟,问答15分钟. (survey一个领域/一篇文章精读/ 一个知识点代码实现)

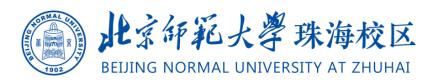
可酌情加沙。/Recommender-System tell-个人形式: Survey/科学论文, 按论文组织合理性, 故事完整性, 思考深度进行评分,

• 附加分

July Madell sithub. com/Madell 书籍、文章、网帖的精华内容,未包含在课程内的surprise项,

Github的使用: https://zhuanlan.zhihu.com/p/94008510

*surprise也是推荐系统的目标



开源数据集

• MovieLens: <u>浅谈推荐系统+3个小时上手python实现(完整代码) - 知乎 (zhihu.com)</u>.

地址: MovieLens | GroupLens.

UCI: 该网站目前维护了436个经典的机器学习、数据挖掘数据集,包含分类、聚类、回归 等问题下的多个数据集。

Kaggle: 全球最大的数据竞赛平台。Kaggle如何入门? - 知乎 (zhihu.com)

• 天池: 阿里旗下数据科学竞赛平台。



报告及论文的参考资料

书籍

- 《深度学习推荐系统》3.8-3.10,第四章,第八章。nder—System
 'Ricci, Lior Rokach, B. Shanira P F. Ricci, Lior Rokach, B. Shapira, Recommender Systems: Handbooks, 2nd edition, Springer, 2015 2.
- M. Zanker, A. Felferning, and Friedrich, Recommender System: An Introduction, Cambridge University Press, 2010 3.
- D. K. Agarwal, B.-Chung Chen, Statistical Methods for Recommender Systems, Cambridge University Press, Feb. 2016 4.

embedding

- Grbovic, Mihajlo, and Haibin Cheng. "Real-Time Personalization Using Embeddings for Search Ranking at Airbnb." Proceedings of the 24th ACM SIGKDD International Conference on Knowledge Discovery & Samp; Data Mining, 2018, pp. 311–320.

 Barkan, Oren, and Noam Koenigstein. "Item? Vec: Newell Land Total Conference on Knowledge Discovery & Samp; Data Mining, 2018, pp. 311–320. 2.
- 3. github. com/Madel



Feature类

- -System Rendle, Steffen. "Factorization Machines." 2010 IEEE International Conference on Data Mining, 2010, pp. 995–1000.
- Guo, H., et al. "DeepFM: An End-to-End Wide & Deep Learning Framework for CTR Prediction." (2018).
- Liu, Bin, et al. "AutoFIS: Automatic Feature Interaction Selection in Factorization Models for Click-Through Rate Prediction." Proceedings of the 26th ACM SIGKDD International Conference on Knowledge Discovery & Samp; Data Mining, 2020, pp. 2636–2645. 3.

时序信息类

- Vaswani, Ashish, et al. "Attention Is All You Need." Proceedings of the 31st International Conference on Neural Information Processing Systems, vol. 30, 2017, pp. 5998–6008.
- Y. Liu, K. Ge, X. Zhang L. Lin, Real-Time Attention Based Look-alike Models for Recommender Systems, https://arxiv.org/pdf/1906.05022
- 3. Qiwei Chen et al, Behavior Sequence Transformer for E-commerce Recommendation In Alibaba, DLP-KDD'19:Proceedings of the 1st International Workshop on Deep Learning Practice for High-Dimensional Sparse Data 2019.
- 3W字长文带你轻松入门视觉transformer 知乎 (zhihu.com) (seg-2-seg, attention, tranfromer) 4.
- Guorui Zou et al, Deep Interest Evolution Network for Click-Through Rate Prediction, AAAI-19 5.
- W. Zhu, D. Tao, X.Cheng, Multi-intereste Network with Dynamic Routing for Recommendation At Tmall,CIKM '19: The 28th ACM International Conference on Information and Knowledge Management, 2019 6.



Multi-objective

- ti-objective

 Ma, X., Zhao, L., Huang, G., Wang, Z., Hu, Z., & Zhu, X., et al. (2018). Entire Space Multi-Task Model: An Effective Approach for Estimating Post-Click Conversion Rate. ACM. ACM.
- Ma, J., Zhe, Z., Yi, X., Chen, J., Hong, L., & Chi, E. H. (2018). Modeling Task Relationships in Multi-task 2. Learning with Multi-gate Mixture-of-Experts. ACM.
- 3. Ruder, Sebastian. "An Overview of Multi-Task Learning in Deep Neural Networks." ArXiv Preprint ArXiv:1706.05098, 2017.
- Lin, X., Chen, H., Pei, C., Sun, F., Xiao, X., Sun, H., Zhang, Y., Ou, W., and Jiang, P. (2019b). A pareto-efficient algorithm for Stephen and Stephen in a commerce research. multiple objective optimization in e-commerce recommendation. In Proceedings of the 13th ACM Conference on Recommender Systems, RecSys'19, pages20–28.
- Milojkovic, Nikola, et al. "Multi-Gradient Descent for Multi-Objective Recommender Systems." ArXiv: Information 5. MadelineMa Retrieval, 2020.

应用类

W. Zhu, D. Tao, X.Cheng, Multi-intereste Network with Dynamic Routing for Recommendation At Tmall, CIKM '19: The 28th ACM International Conference on Information and Knowledge Management, 2019

graph convolutional network有什么比较好的应用task剂如知乎 (zhihu.com) **译释器**

SHAP: Python的可解释机器学习库 - 知乎 (zhihu.com)

强化学习基础知识

强化学习入门 第一讲 MDP - 知乎 (zhihu.com)

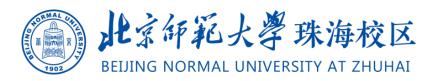
强化学习(二)马尔科夫决策过程(MDP)-刘建平Pinard-博客园(cnblogs.com)

ICML 2019 | 强化学习用于推荐系统, 蚂蚁金服提出生成对抗用户模型 - 知乎 (zhihu.com)

SuttonBartoIPRLBook2ndEd.pdf (stanford.edu)

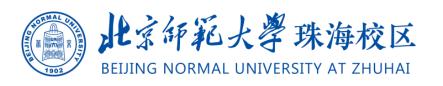
2021顶会文章

KDD2021推荐系统, 计算广告论文汇总 - 知乎 (zhihu.com)



We can read of things that happened 5,000 years ago in the Near East, where people first learned to write.





TelineMa/Recommender-System 北京师范大学数学研究中心.

办公室: 励教楼B406

本科应用数学学士学位,计算机科学与技术博士学位.

目前研究方向主要在多目标学习,推荐系统,可解释机器学习在推荐和医疗方向的应用。推荐系统课件及资料之间。

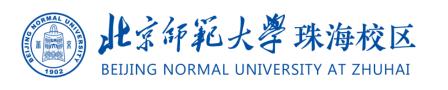
推荐系统课件及资料github: https://github.com/MadelineMa/Recommender-System

github. com/MadelineMa/



学生简介elineMa/Recommender-System
github.com/MaplelineMa/Recommender-System

Ma/Recommen We can read of things that happened 5,000 years ago in the Near East, where people first the learned to write.



本周作业

- 1. 多刷淘宝 VS 京东 VS 得物 VS B站,与系统交互兴趣,感受推荐力度。
- 2. 自由组队,依参考文献但不限于参考文献,选择和推荐系统相关的课题准备Report,给出队员名 单和选题题目,方便教师排序。(3周内)
- 3. 依据兴趣选择继续学习或退课。

