

# ITM-Rec: An Open Data Set for Educational Recommender Systems

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### **Abstract**

With the development of recommender systems (RS), several promising systems have emerged, such as context-aware RS, multi-criteria RS, and group RS. However, the education domain may not benefit from these developments due to missing information, such as contexts and multiple criteria, in educational data sets. In this paper, we announce and release an open data set for educational recommender systems. This data set includes not only traditional rating entries, but also enriched information, e.g., contexts, user preferences in multiple criteria, group compositions and preferences, etc. It provides a testbed and enables more opportunities to develop and examine various educational recommender systems.

## **Background & Data Collections**

The data set was collected by using Qualtrics from 2017 to 2022.

- Subjects: graduate students at the ITM department in Illinois Tech.
- Courses: database (DB), data analytics (DA), and data science (DS).
- Items: topics for final projects, e.g., Kaggle data sets in DA & DS courses.
- Survey 1: data collection of individual preferences
- Survey 2: data collection of group preferences

### **Survey 1: individual preferences**

- Student demographic info, e.g., age, gender, marriage status
- Selected topics/data: at least 3 liked and 3 disliked ones for final projects
- Overall rating: 3-5 for liked items, 1-3 for disliked items
- Multi-criteria ratings: ratings on App, Data, Ease
  - App: how they like the application domain of the data sets/topics
  - Data: how they like the data processing or storage
- Ease: how they like the overall ease if using it for final projects
- Example of rating collections: see table below

### Survey 2: group preferences

- Students may eventually work on projects on their own, or via a teamwork.
- For students with teamwork, students must fill out survey 2 by a group discussions. Each group filled a single copy of survey 2.
- Questions in survey 2 are similar to the ones in survey 1. There are only questions related to rating collections.
- Eventually, we have group compositions, group preferences and individual preferences of each group member in our data sets.

## **Data & Descriptions**

The whole data set is composed of five major files:

- users.csv columns: UserID, Gender, Age, Married description: Meta data about students
- items.csv
  columns: Item, Title, URL, Descriptions (texts or texts crawled from URL)
  description: Meta data about the topics of projects
- ratings.csv columns: UserID, Item, Rating, App, Data, Ease, Class, Semester, Lockdown description: Students' individual ratings on items, including the overall rating and multi-criteria ratings, as well as three contextual variables, e.g., class, semester, lockdown periods (pre, dur, pos) statistics: 5,230 ratings given by 476 students on 70 items, sparsity: 84.30%
- group.csv columns: GroupID, UserID description: the compositions of groups

Rating

UserID

1173

ItemID

28

group\_size.csv
columns: GroupID, Size
description: the number of students in each group
statistics: group size: {2: 88, 3: 42, 4: 9, 5: 4}

App

group\_ratings.csv
columns: GroupID, Item, Rating, App, Data, Ease, Class, Semester, Lockdown description: Ratings on items given by groups, rather than individuals statistics: 1,117 ratings given by 143 groups on 70 items, sparsity: 88.84%

Data

Ease

Course

1175 		41 	5 4 		4		4 DS		Spring 		POS 	
Rating	1	0.8	0.72	0.58	- 1.0 - 0.9	Rating	1	0.79	0.68	0.57	- 1. - 0.	
Арр	0.8	1	0.78	0.63	-0.8	Арр	0.79	1	0.74	0.65	- 0.	
Data	0.72	0.78	1	0.66	- 0.7	Data	0.68	0.74	1	0.71	- 0.	
Ease	0.58	0.63	0.66	1	- 0.6	Ease	0.57	0.65	0.71	1	-0.	
	Rating	Арр	Data	Ease			Rating	Арр	Data	Ease		
C	Correlati	ions on I	Individu	al Rating	S		Correl	ations o	n Group	Rating	S	

# **Example of Usage**

The data set can be utilized for educational analysis and educational recommender systems (EdRec). Below are some examples of EdRec.

- Traditional EdRec with side information [1] user demographic info: age, gender, marriage status item features: textual descriptions
- Context-Aware EdRec [2] context variables: course, semester hint: adapter recommendations to contextual situations
- Multi-Criteria EdRec [3] criteria: app, data, ease hint: enhance recommendations from multi-criteria preferences
- Group EdRec [4]
  necessary info: group compositions and group preferences
  hint: evaluate different group aggregation and recommendation strategies
- Multi-Objective EdRec [5]
  hint: balance multiple objectives in educational scenarios
  examples: balance accuracy, diversity and novelty in recommendations
- Integrating of Context and Multi-Criteria Preferences [6] hint: enhance EdRec by fusing context and multi-criteria info together
- Integrating of Group and Multi-Criteria Preferences [7] hint: enhance Group EdRec by fusing multi-criteria preferences

### **Open Discussions**

- Any other ways to take advantage of this data set?
- Any other suggestions in future data collections?
- Any similar open data sets in education?
- Any potential workshops to host a data or recommendation challenge?

# Release & Downloads Kaggle GitHub SCAN ME



# References

L. Cantas, O. C. P. Datisaria, L. C. (Ed.). (2011). Educational Decommander Systems and Tochnologies, Drastices and Challenges, ICI Clabe

5. Zheng, Y., & Wang, D. X. (2022). A survey of recommender systems with multi-objective optimization. Neurocomputing, 474, 141-153.

- 1. Santos, O. C., & Boticario, J. G. (Ed.). (2011). Educational Recommender Systems and Technologies: Practices and Challenges, IGI Global.

  2. Zheng, Y. and Mobasher, B. (2018). Context-Aware Recommendations. In Book: Collaborative Recommendations: Algorithms, Practical Challenges and Applications, World Scientific Publishing, 173-202
- 3. Zheng, Y., & Wang, D. (2022). Multi-Criteria Ranking: Next Generation of Multi-Criteria Recommendation Framework. IEEE Access, 10, 90715-90725.
- 4. Zheng, Y. (2018). Exploring user roles in group recommendations: A learning approach. In Adjunct Publication of the 26th Conference on User Modeling, Adaptation and Personalization.
- 6. Zheng, Y., Shekhar, S., Jose, A. A., & Rai, S. K. (2019). Integrating context-awareness and multi-criteria decision making in educational learning. In ACM/SIGAPP Symposium on Applied Computing.
  7. Zheng, Y. (2019). Educational group recommendations by learning group expectations. In IEEE TALE.

Semester Lockdown

Fall

PRE