

## Data Analytics

### Project Presentation

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## Abstract and Scope

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- Overwhelmingly large variety of products online
- No single model exists that can simultaneously optimize multiple objectives
- Helping online retailers select more relevant items from a vast range to recommend to their customers based on their real-time behavior
- Improving recommendations to ensure navigating through seemingly endless options is more effortless and engaging for shoppers.
- The recommender system along with accurately predicting a user's buying habits, must also include some form of novelty and serendipity in its predictions in order to expand the user's taste as a job of a recommender system is not only to recommend the right items but also keep the user engaged.

## Goal of the Project

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Building a multi-objective recommender system based on previous events in a user session to predict e-commerce clicks, cart additions, and orders

## Project Demonstration

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Exhibit the demonstration of the complete project.

## Implementation Details

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### Module 1: Exploratory Data Analysis (EDA)

- Data Visualisation and Description using Python and its libraries including numpy, pandas, matplotlib, seaborn
- Comparisons drawn between number of various events (clicks, carts, orders)
- Finding of the top 20 products most interacted with in terms of various events.
- Visualisation of sessions

## Implementation Details

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### Module 2: Model Building- Model 1 using Word2Vec

- First model was built using Word2Vec to learn word associations from a large corpus of text
- Corpus created for clicks, carts and orders

## Implementation Details

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### Module 3: Model Building- Model 2 Baseline Model

- The task is to predict the next aid clicked after the session truncation, as well as the the remaining aids that are added to carts and orders
- For each session in the test data, we have to predict the aid values for each type that occur after the last timestamp

## Results and Discussion

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We aimed to build a multi-objective recommender system based on previous events in a user session to predict e-commerce clicks, cart additions, and orders

Our recommender system must maximise all three: accuracy, novelty and serendipity and thus is a multi objective optimisation problem.

According to our results, we are able to predict possible events for each user



## Conclusion

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- There is an emerging demand in multi-objective optimization recently in RecSys, especially in the area of multi-stakeholder and multi-task recommender systems
- The multi-objective recommendation is an application of MOO(MULTI-OBJECT OPTIMIZATION) in the area of recommender systems.

## Conclusion

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- The aim of this project was to predict e-commerce clicks, cart additions, and orders by building a multi-objective recommender system based on previous events in a user session.
- Current recommender systems consist of various models with different approaches, ranging from simple matrix factorization to a transformer-type deep neural network. However, no single model exists that can simultaneously optimize multiple objectives. we tried to build a single entry to predict click-through, add-to-cart, and conversion rates based on previous same-session events.

### Model Building- Model 3 Candidate reranking model

In the final stage of a recommendation system, the system can re-rank the candidates to consider additional criteria or constraints. One re-ranking approach is to use filters that remove some candidates.

#### Step 1 - Generate Candidates

For each test user, we generate possible choices, i.e. candidates. we generate candidates from 5 sources: user history, most popular events, co visitation matrix of events with type weighting, buy2buy and time weighting

### Model 3 Candidate reranking model

- Step 2 - ReRank and Choose 20

Given the list of candidates, we must select 20 to be our predictions. We can do this with a set of handcrafted rules. We can improve our predictions by training an XGBoost model to select for us.

## Future Work

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### Suggested Models:

- Transformers
- Particle Swarm Optimisation

## References

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Thank  
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