

Slate Recommendation Systems

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Slate
Recommendation
Systems

Combinatorial Action Spaces

Long Term Engagement

Real Time Recommendation

Problem Objective

Utilize slate-based reinforcement learning techniques to optimize the selection of reading material on a 3x3 clickable grid.

Problem Space

Nine cards are displayed out of 350 options on a reading grid

Newer choices are often added, older ones aren't removed

Feature data is available on users, predicted to be single-use

Correlation expected between different reading items.

Placement of items also expected to play a role in output

There may be seasonality and trends in the reading material's popularity.

Current Approach

Random
Exploration

Greedy
Most visited recently

Rule-Driven
Feature Data

Linear Contextual Bandit
User-feature based

User Choice Model

- The primary motivation to not consider each item as one action and then simply work on the choice of the top 9 is to consider interaction effects between items.
- To this effect, we create a user-choice model that will work on identifying probability of selection of every item from a given slate.
- This model can be built learning from user interactions or by a heuristic/rule-based method.
- The former would involve significant investment into it since it's own accuracy would have a significant impact on the results of our algorithm.
- The latter would allow us to focus on a simpler model and invest more time into learning from the same.

SlateQ Recommendation Systems



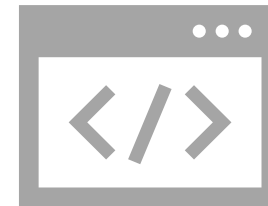
Q Learning

Assumes a prepared user choice model

Generates LTVs for each item

Uses TD Learning to learn LTVs

Uses a DQN with Experience Replay



Slate Optimisation

Mixed Integer Programming offline

Uses k-best or sequentially greedy methods
online

Proposed Next Steps

1

Obtaining and
cleaning data

2

Creating and
understanding
user features

3

Creating a User
Choice Model

4

Implementing
the SlateQ
Algorithm

5

Factoring in
placement
metrics

6

Implementing
trends and
seasonality