INTEGRATING BEHAVIORAL DATA FOR TAILORED OUTFIT RECOMMENDATIONS

Personalization became crucial for user engagement and retention as the online fashion industry grew in popularity. Recommender systems in e-commerce are generally built using collaborative filtering, content-based filtering, or hybrid approaches. Collaborative filtering identifies user similarities based on shared preferences [1], while content-based filtering leverages item features to recommend similar products [2]. However, traditional recommendation methods face limitations. This study aims to improve recommendation quality on a fashion rental platform by integrating behavioural data (e.g., browsing history, and rental frequency). Behavioral data provides insights into users' implicit preferences [3]. Integrating behavioral data enhances the effectiveness of recommendation systems by capturing a fuller picture of each user's unique preferences and needs.

Table 1. Behavioral Data Metrics

Metrics	Description
Browsing	Users frequently viewing specific styles, brands, or
history.	colours will likely prefer those attributes. The
-	recommendation system can identify user visual
	preference trends by analyzing browsing patterns. Even
	if they have not rented an item, it helps to make better
	suggestions for future rentals.
Rental	High-frequency renters may seek variety, while
frequency.	low-frequency renters may prefer staple pieces.
	Recommendations for frequent renters might prioritize
	diverse styles, while infrequent renters could receive
	suggestions for versatile, seasonally relevant items.
Rental	Users who keep items longer usually prefer classic,
duration	timeless styles, while other customers may seek trendy
	or event-specific outfits. Depending on usage patterns,
	analyzing rental duration can help recommend timeless
	basics or fast-fashion items.
Ratings and	Positive reviews for specific items or styles mean
reviews	user satisfaction and could prompt similar
	recommendations. Conversely, negative reviews could be factored in to avoid recommending similar items.
The pattern	Repeat rentals suggest loyalty to specific styles or
<i>The pattern</i> of renting the	fits. The recommender algorithms should highlight
same or similar	similar items or newer versions of rented pieces,
items multiple	assuming the user has a strong preference
times	ussuming the user has a strong preference
Time of	This information can reveal peak shopping times
interaction	and seasonal preferences; for example, users who
	browse or rent more during specific seasons might
	appreciate recommendations aligned with seasonal
	trends or weather-appropriate items
Cart	Cart abandonment data offers insight into what
abandonment	users find appealing but hesitate to rent, perhaps due to
	price or uncertainty about fit. A recommendation
	algorithm could re-surface these items when they go on
	sale or show similar products at lower prices
User	In-depth sessions might indicate a user's interest in
session duration	exploring multiple options before choosing. Diverse and
<i>and depth</i> , including the	exploratory recommendations can be provided for such users, while shorter sessions might suggest a preference
including the number of pages	for direct, targeted recommendations.
viewed	for uncet, targeteu recommentations.
Outfit	Users who rent outfits in sets (e.g., pairing tops with
pairing patterns	bottoms or adding accessories) may benefit from
pairing purcents	bundled recommendations or outfit suggestions that fit
	their pairing patterns.
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This study confirms that integrating behavioural data through a hybrid recommendation model can significantly improve the quality of outfit recommendations on a fashion rental platform. Incorporating behavioural profiles allows for more personalized user experiences, which can be a competitive advantage in the fashion rental market.

References

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