

Amazon Magazine Subscription Recommendation System

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The problem

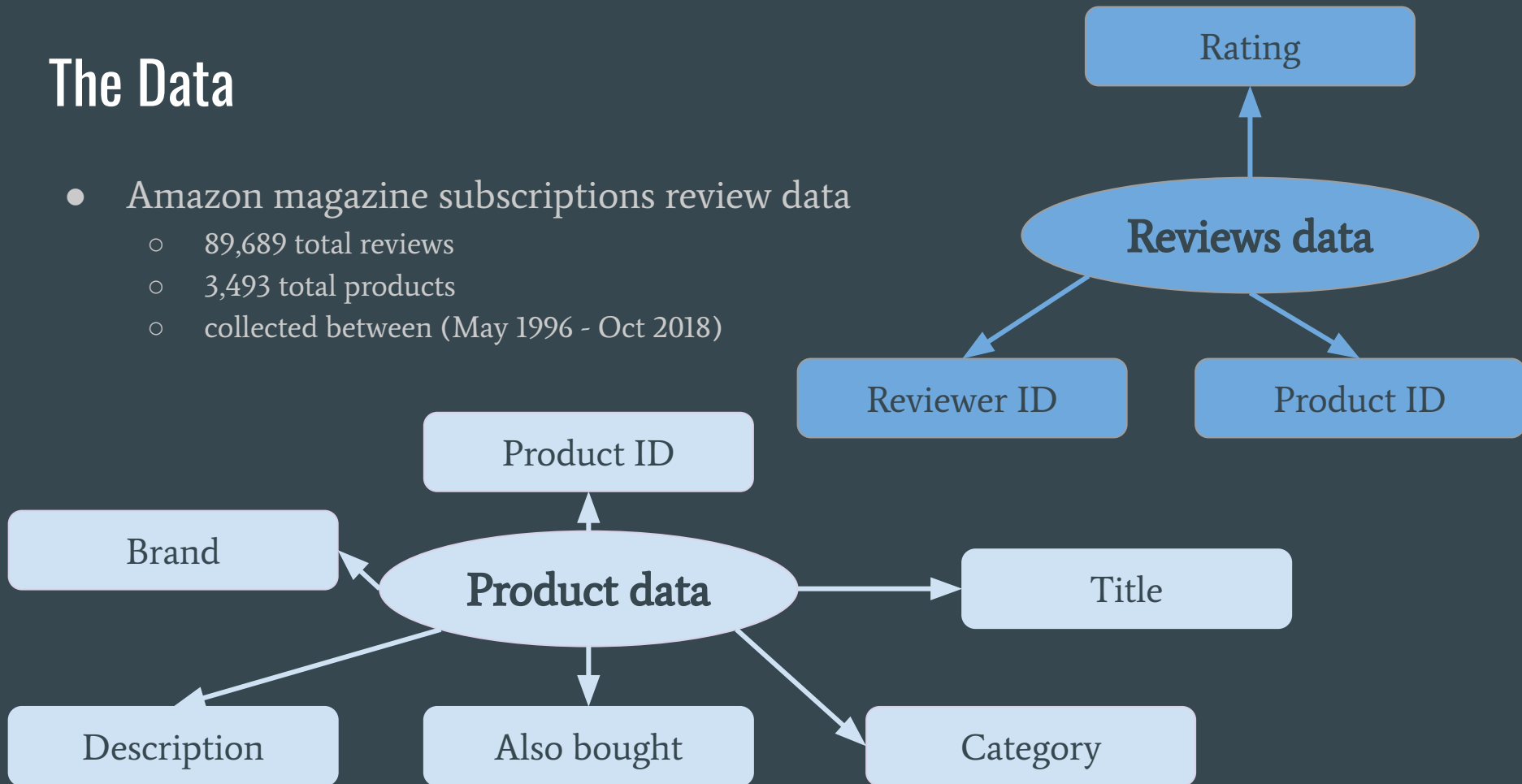
- Information overload makes it difficult for customers to find the right product
- There is an opportunity to increase sales by putting the most relevant products in front of the right customers

Our solution

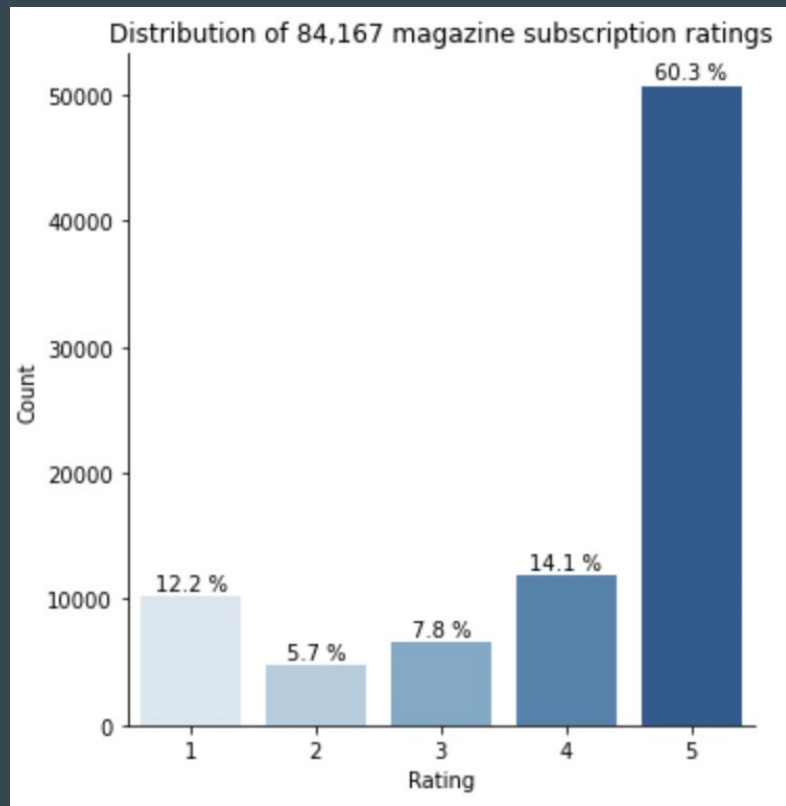
- Personalize recommendations based on:
 - Similar users
 - Similar products
 - Most popular products
- Adopt a data-driven method to increase sales and reduce churn

The Data

- Amazon magazine subscriptions review data
 - 89,689 total reviews
 - 3,493 total products
 - collected between (May 1996 - Oct 2018)

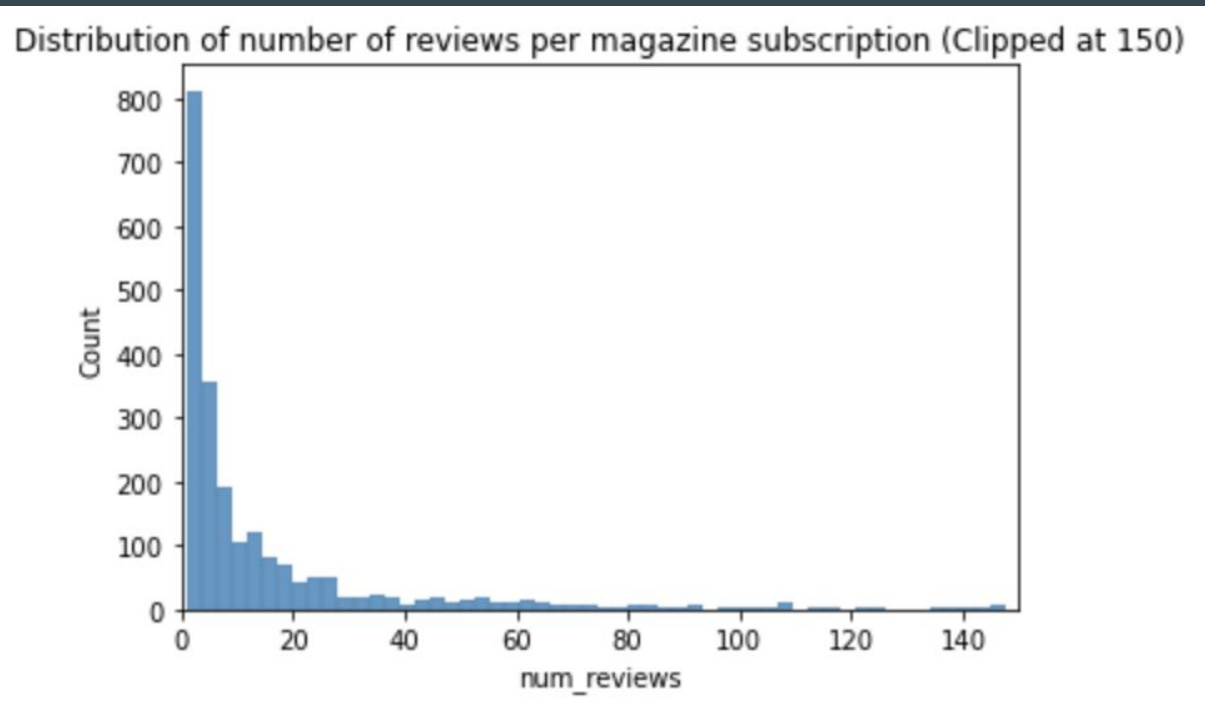


Data Exploration - Distribution of Ratings



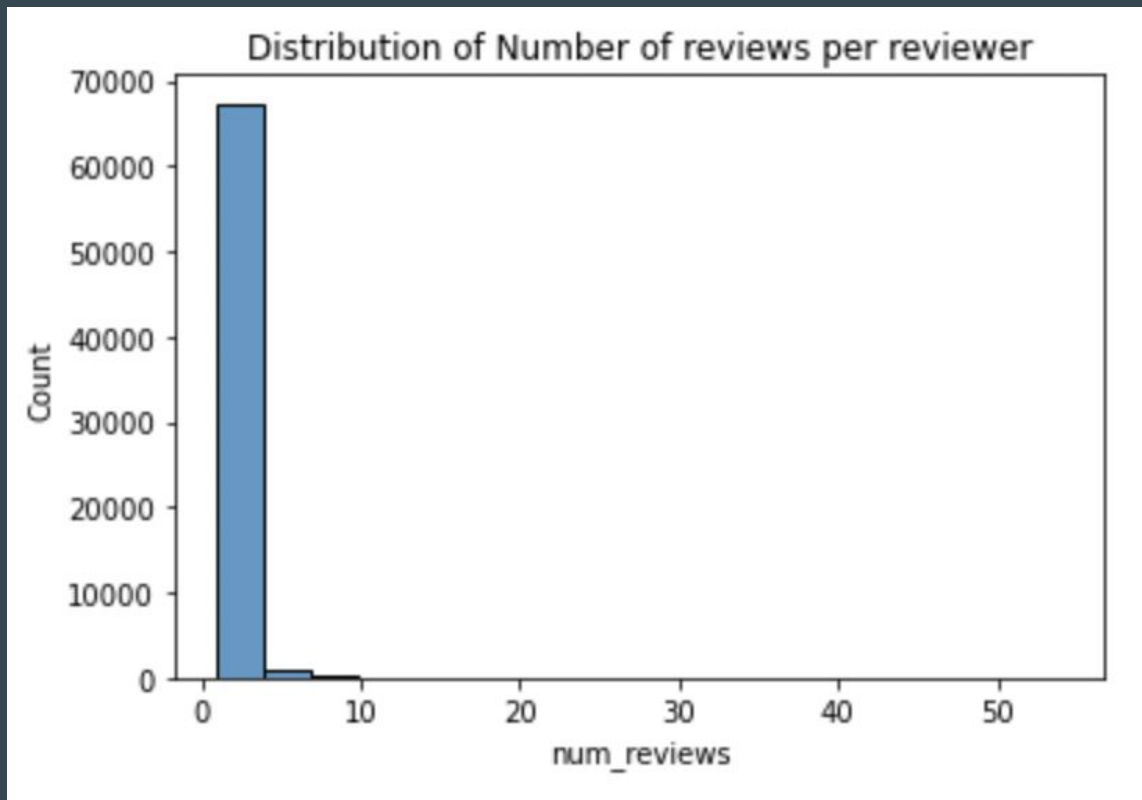
- 60.3% of magazine subscriptions received 5 star ratings
- Very few magazine subscriptions received 2 & 3 star ratings

Data Exploration - Distribution of Reviews



- Most magazine subscriptions received only 1 rating
- The most rated magazine received 1718 reviews

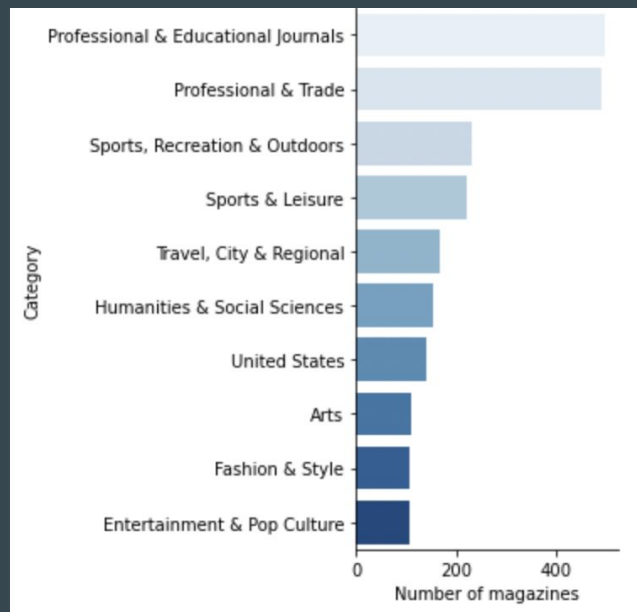
Data Exploration - Distribution of Reviewers



- Most reviewers only gave 1 review
- The most productive reviewer gave 54 reviews

Data Exploration - Magazine Categories

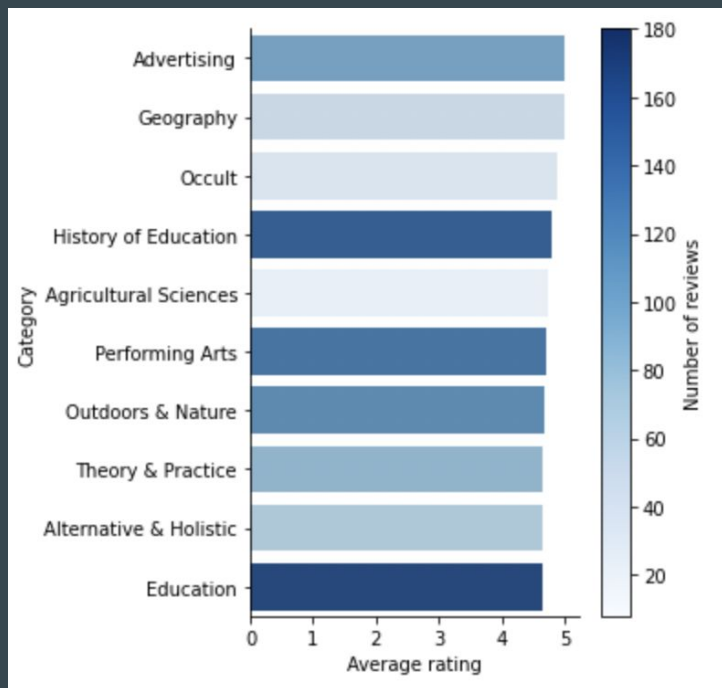
Top categories by number of magazines



- Most magazines are categorized as Professional & Education journals (497 in total)

Data Exploration - Magazine Categories

Top categories by average rating



- Magazines categorized as Advertising has the highest average rating (5-stars)
- 'Education' and 'History of Education' magazines received the most reviews

Data Exploration - Most loved magazine subscriptions

1. Mystery Scene



2. House & Home



3. Gray's Sporting Journal



4. B.O.S.S



5. Dirt Bike



Data Exploration - Most disliked magazine subscriptions

1. American Baby



2. Elle Decor - Italian ed



3. Variety



4. Elle - French ed



5. V Magazine



Recommendation Systems

1. **Collaborative Filtering** using Alternating Least Squares (PySpark)
 - Test RMSE: 1.215
2. **Collaborative Filtering** using various prediction algorithms (Python)

Algorithm	KNN Baseline	SVD	Baseline Only	SVD pp	Co- Clustering	KNN With Means	Slope One	KNN With ZScore	KNN Basic	NMF	Normal Predictor
Test RMSE	1.349341	1.351881	1.362914	1.368660	1.382874	1.384150	1.387706	1.388441	1.399252	1.403964	1.800689

3. **Content-based Filtering** using cosine similarity

Recommendation Systems - Best Models

1. Collaborative filtering using Alternating Least Squares (PySpark)
 - Test RMSE: 1.215
 - Params:
 - Rank = 1, MaxIter = 1, RegParam = 1
2. Content-based filtering using cosine similarity

Case study: Amazon Customer

Maximum PC



- Technology
- Computers & Internet

Vanity Fair



- Professional & Educational Journals
- Professional & Trade
- Entertainment & Media

PC World



- Technology
- Computers & Internet

Details



- Fashion & Style Men

Wired



- Science
- History & Nature
- Technology

GQ



- Fashion & Style Men

Bon Appétit



- Cooking
- Food & Wine
- Recipes & Techniques

Xbox



- Literary
- Sci-Fi & Mystery
- Science fiction & fantasy
- Fantasy

Case study: Amazon Customer - Summary

- Male
- Interests:
 - Men's fashion
 - Cooking
 - Technology
 - Sci-fi
 - History & nature
 - Science
 - Literary
- Reviewed 5 Conde Nast magazines, 4 of which received 5-stars
- Reviewed 2 Future US, Inc. magazines, one received 1-star whilst the other received 5-stars

Recommendations - based on similar reviewers (ALS)

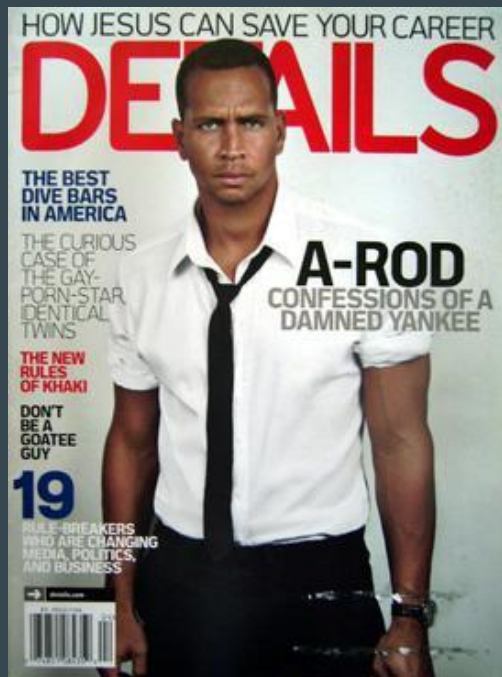
	Title	Brand	Category	Summary
1	Gourmet News	Oser Communications Group Inc	<ul style="list-style-type: none">• Cooking• Food & Wine• Reference	<ul style="list-style-type: none">• Gourmet industry• New products• Trade news
2	Diapason	Scranton Gillette Commun Inc	<ul style="list-style-type: none">• Professional & Educational Journals• Professional & Trade• Entertainment & Media• Music	<ul style="list-style-type: none">• Organ, harpsichord & church music
3	Architectural Digest	Conde Nast	<ul style="list-style-type: none">• Arts• Music & Photography	<ul style="list-style-type: none">• Features work of top architects & interior decorators
4	Allure	Conde Nast	<ul style="list-style-type: none">• Fashion & Style	<ul style="list-style-type: none">• Beauty expert advice & tips
5	Professional Artist	Turnstile Publ	<ul style="list-style-type: none">• Music & Photography,• Art & Art History	<ul style="list-style-type: none">• Guide artists on their journey toward making a living with their art

Do these recommendations make sense?

	Title	Matching criterias	Possible reasoning
1	Gourmet News	<ul style="list-style-type: none">• Categories:<ul style="list-style-type: none">◦ Cooking◦ Food & Wine	
2	Diapason		<ul style="list-style-type: none">• A high proportion of the user's demographic may be interested in this magazine
3	Architectural Digest	<ul style="list-style-type: none">• Brand: Conde Nast	
4	Allure	<ul style="list-style-type: none">• Brand: Conde Nast	<ul style="list-style-type: none">• Fashion & Style Men is a subset of Fashion & Style
5	Professional Artist		<ul style="list-style-type: none">• Art & Art History is similar to History & Nature

Recommendations - based on similar magazines

Which magazines are most similar to 'Details' ?



- Fashion & Style, Men
- Conde Nast Publications
- “Hottest fashion, celebrities, movies, music, ideas, technology & issues of the day”

Recommendations - based on similar magazines (Details magazine)

	Title	Brand	Category	Summary
1	Vanity Fair	Conde Nast Publications	<ul style="list-style-type: none">• Professional & Educational Journals• Professional & Trade• Entertainment & Media	<ul style="list-style-type: none">• People, personalities & power
2	Cowboys & Indians	Cowboys & Indians	<ul style="list-style-type: none">• Travel• City & Regional• United States• South	<ul style="list-style-type: none">• Luxury western lifestyle• Western & American Indian art, interiors, fashion and jewelry
3	Esquire	Reader's Digest Association	<ul style="list-style-type: none">• Fashion & Style Men	<ul style="list-style-type: none">• Men's lifestyle• Style & clothes, cars, culture & entertainment and advice on money
4	Coastal Living	Meredith Corporation	<ul style="list-style-type: none">• Home & Garden• Design & Decoration	<ul style="list-style-type: none">• Homes & destinations on coastlines of US, Canada & Carribean• Recipes, coastal news, products & events
5	Glamour	Conde Nast Publications	<ul style="list-style-type: none">• Fashion & Style Women	<ul style="list-style-type: none">• Hair & Beauty• Relationship & sex questions• Monthly horoscopes• Health & diet

Do these recommendations make sense?

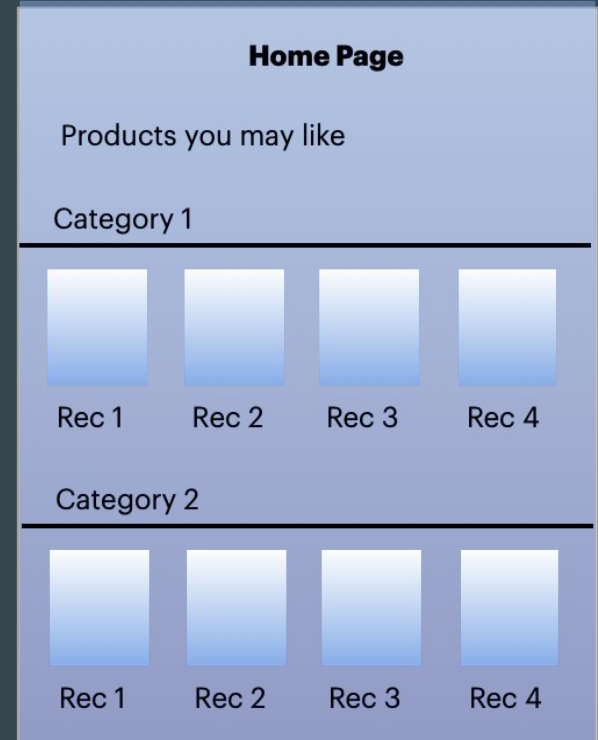
	Title	Matching criterias	Possible reasoning
1	Vanity Fair	<ul style="list-style-type: none">• Brand: Conde Nast	<ul style="list-style-type: none">• Celebrities
2	Cowboys & Indians		<ul style="list-style-type: none">• A type of Men's Lifestyle
3	Esquire	<ul style="list-style-type: none">• Categories:<ul style="list-style-type: none">○ Fashion & Style Men	
4	Coastal Living		<ul style="list-style-type: none">• A type of Men's lifestyle
5	Glamour	<ul style="list-style-type: none">• Brand: Conde Nast	<ul style="list-style-type: none">• Lifestyle content

Conclusion

- Best models:
 - Collaborative filtering using Alternating Least Squares (PySpark)
 - Test RMSE: 1.215
 - Content-based filtering using cosine similarity
- Recommendation systems can help increase sales & reduce churn by:
 - Displaying the most relevant items that users are interested in
 - Helping users discover new items

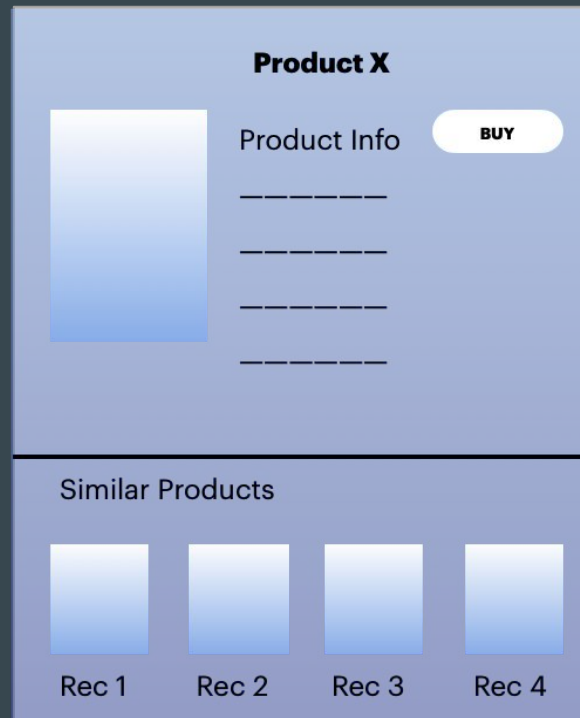
Recommendations - Collaborative Filtering Recommendation System

- Display collaborative filtering recommendations on a customer's home page



Recommendations - Content-Based Filtering Recommendation System

- Display content-based filtering recommendations on product pages



Recommendations - Cold Start Problem

- Cold Start Problem:
 - New users don't have any ratings
 - Recommendation systems will not work
- Solution:
 - Show new users the most popular items (generated from exploratory data analysis)



Future Scope

- Use image recognition to include magazine covers
- Sentiment analysis on review text
- Use A/B testing to determine which recommendation system works best/ the best places to display recommendations
- Additional dataset:
 - User demographic
 - Entire Amazon Reviews dataset (all categories)