



UNIVERSITAT DE
BARCELONA



Master on Foundations of Data Science



Recommender Systems

Course Presentation

Santi Seguí | 2018-2019



Associate Professor at the Department of Mathematics and Computer Science
from the University of Barcelona.

PhD in 2011 on Computer Vision and Machine Learning
at the University of Barcelona.

Graduate on Computer Science in 2007

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Why recommender systems?

“A lot of times, people doesn’t know what they want until you show it to them...”

–Steve Jobs, 1997

“People read around 10 MB worth of material a day, hear 400MB a day, and see 1MB of information every second”

–The Economist, November 2006

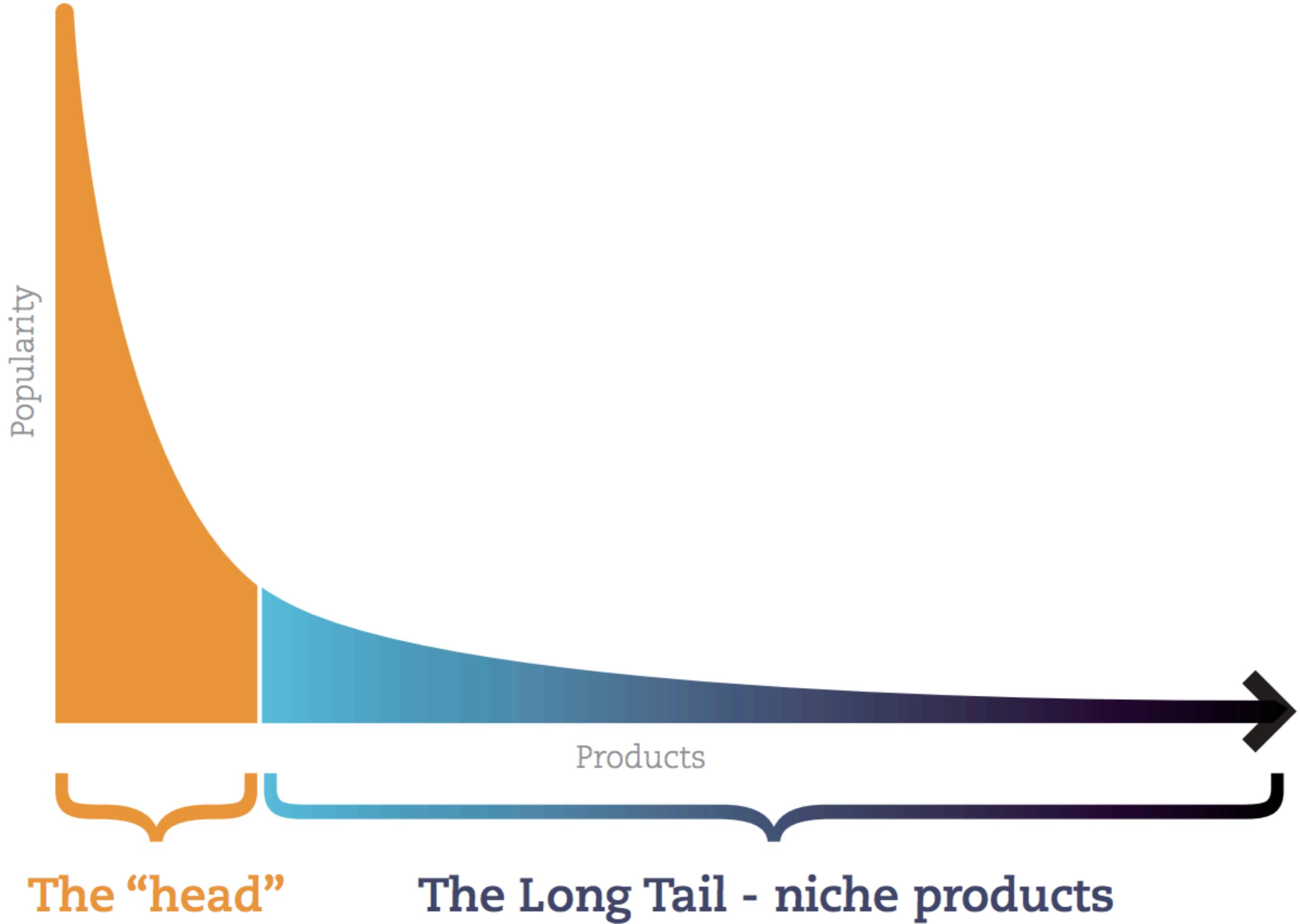
“We are leaving the information age, and entering into the recommendation age”

–Chris Anderson, from the Long Tail Book - 2008

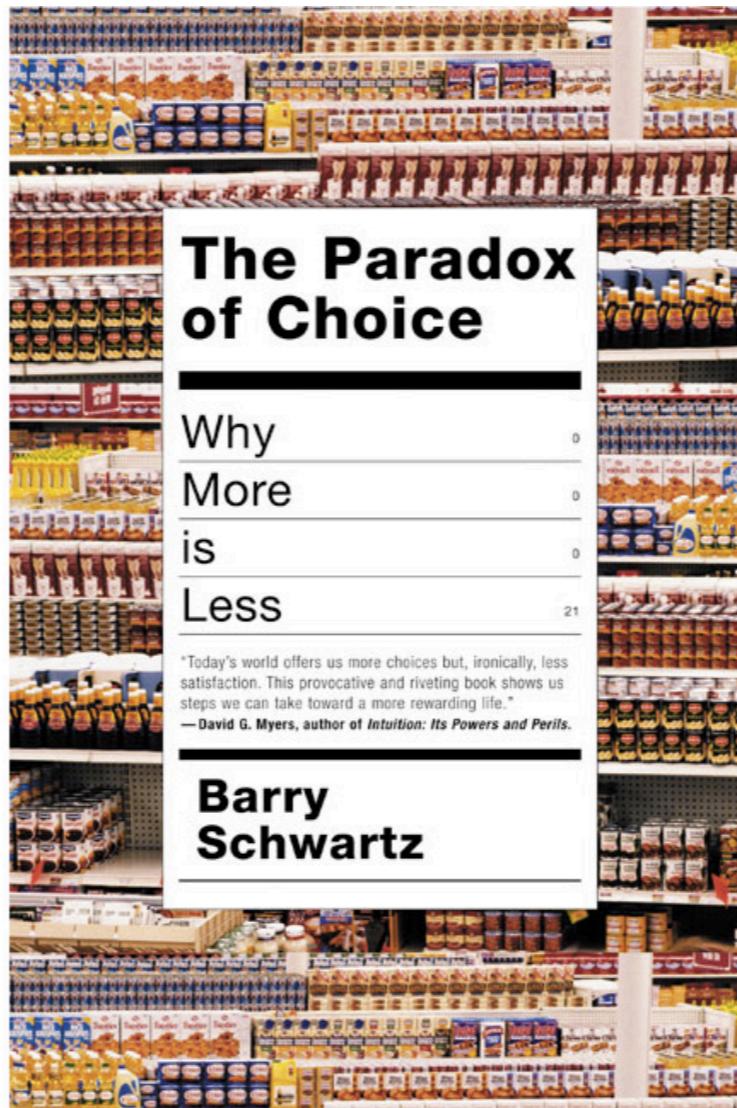
“In 2015, consumption will raise
to 74GB a day”

—UCSD Study, 2014





Information overload



You go to the doctor and because you are sick and:

Doctor: we could do *A*, or we could do *B*

A has these benefits, and these risks

MOST OF THE TIME

B has these benefits, and these risks

What do you want to do?

WE DO NOT WANT

Doctor: *A* has these benefits, and these risks

B has these benefits, and these risks

TO CHOOSE

What do you want to do?

You: Mmm, If you were me, Doc, What would you do?

Doctor: But I'm not you

What are the recommender systems for?

System designed to recommend **to me**
something **I** may like!.

What are the recommender systems for?

Recommender systems are tools designed for interacting with **large and complex** information spaces and **providing information or items** that are likely to be of **interest to the user, in an automated fashion.**



Recommender Systems Master on Foundations of Data Science



Goal of the course:

Learn the basics on recommender systems, and finally become an expert on this topic

What do we expect from the students?
Active participation in class
Some work at home

Course Agenda - 2018

Labs (Thursday)	
14/02	Introduction to Recommender Systems
21/02	Non-Personalized Recommenders
28/03	Collaborative-Based Recommenders
07/03	Collaborative-Based Recommenders
14/03	Dimensionality Reduction for Recommender Systems
21/03	Content-Based Recommender Systems
28/03	Music Recommender Systems
04/04	Evaluation Recommender Systems
11/04	MATEFEST
19/04	Easter BREAK
25/04	Graph-Based Models
02/05	Deep Learning Models
09/05	Context Based Models
16/05	Group Based Models / Knowledge Recommendations
23/05	Current Practices in Industry and Research
30/05	Exam

Course Evaluation

Assignment #1 : 25% - Collaborative Rec Systems

Assignment #2: 25% - Graph / DL Rec Systems

Presentation #3: 25%

Final Exam : 25%

Who uses Recommender systems?

Who uses Recommender systems?



Recommender systems are everywhere

A bit of History

- 1985 - First ideas of recommender systems were mentioned
- 1992 - Tapestry by Xerox Palo Alto
 - First system designed by collaborative filtering
- 1994 - GroupLens
 - First recommender system using rating data
- 1997 - MovieLens
 - First movie recommender system

Prizes were based on improvement over Netflix's own algorithm, called Cinematch, or the previous year's score if a team has made improvement beyond a certain threshold. A trivial algorithm that predicts for each movie in the quiz set its average grade from the training data produces an RMSE of 1.0540. Cinematch uses "straightforward statistical linear models with a lot of data conditioning"





Watch Instantly

Browse DVDs

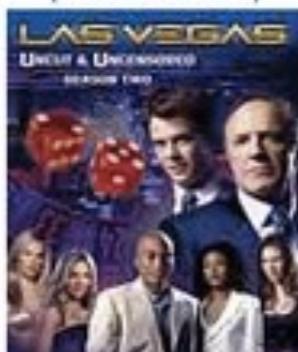
Your Queue

Movies You'll ❤️

Congratulations! Movies we think You will ❤️

Add movies to your Queue, or Rate ones you've seen for even better suggestions.

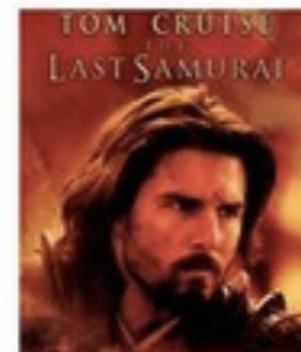
Spider-Man 3

[Add](#)Las Vegas: Season 2
(6-Disc Series)

300

[Add](#)

The Last Samurai



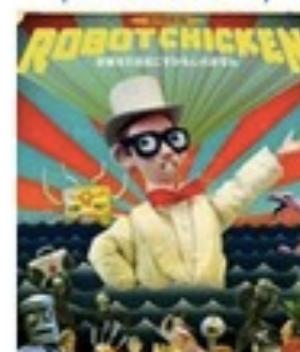
The Rundown

[Add](#)

Bad Boys II

[Add](#)

Star Wars: Episode III

Robot Chicken: Season 3
(2-Disc Series)

IMDb Charts

Top Rated Movies

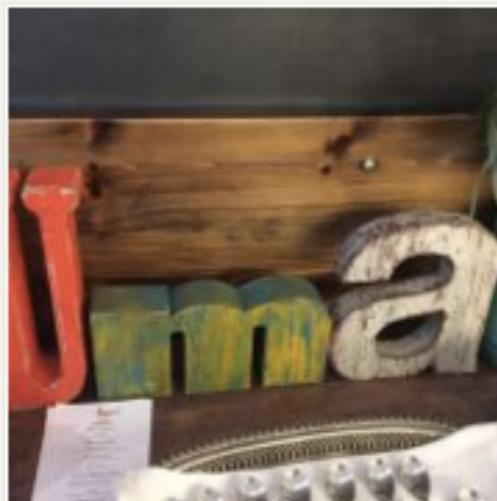
Top 250 as voted by IMDb Users

Showing 250 Titles

Sort by: Ranking



Rank & Title	IMDb Rating	Your Rating	
1. Cadena perpetua (1994)	★ 9,2	☆	
2. El padrí (1972)	★ 9,2	☆	
3. El padrí II (1974)	★ 9,0	☆	
4. El caballero oscuro (2008)	★ 8,9	☆	
5. Pulp Fiction (1994)	★ 8,9	☆	
6. La llista de Schindler (1993)	★ 8,9	☆	
7. 12 hombres sin piedad (1957)	★ 8,9	☆	
8. El senyor dels anells: El retorn del rei (2003)	★ 8,9	☆	



Uma

1 de 7.365 Restaurantes en Barcelona

●●●●● 227 opiniones

"Probarlo" 08/01/2016

"Creatividad en estado puro" 05/01/2016

Precio: 65 € - 75 € | Mapa | Fotos de los visitantes (345)

Cocina:

[Internacional](#)

[Mediterránea](#)

[Fusión](#)

[Española](#)



Tast-Ller

2 de 7.365 Restaurantes en Barcelona

●●●●● 324 opiniones

"COMPARTIENDOLO CON AMIGOS" 13/01/2016

"Excelente" 15/12/2015

Precio: 50 € - 60 € | Mapa | Fotos de los visitantes (312)

Cocina:

[Delicatessen](#)



Fulla d'Ostra

3 de 7.365 Restaurantes en Barcelona

●●●●● 166 opiniones

"Exquisito" 30/12/2015

"Excelente menú" 23/12/2015

Precio: 80 € - 100 € | Mapa | Fotos de los visitantes (75)

Cocina:

[Fusión](#)

[Mediterránea](#)

[Española](#)



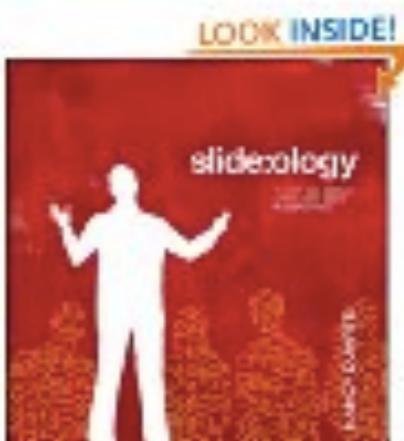
Informal

4 de 7.365 Restaurantes en Barcelona

●●●●● 215 opiniones



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slide:ology: The Art and
Science of Creating
Great Presentations... by Nancy Duarte
 (98)
\$23.09

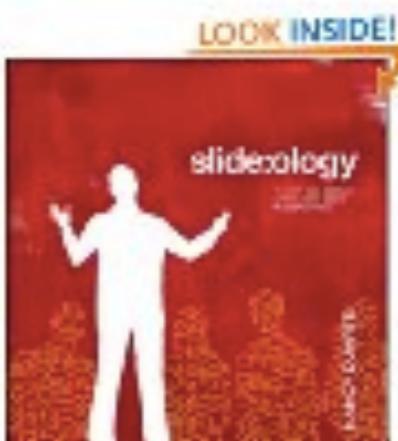


The Naked Presenter:
Delivering Powerful
Presentations... by Garr
Reynolds
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Simplifying Presentations... by Garr Reynolds
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slide:ology: The Art and
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Great Presentations... by Nancy Duarte
 (98)
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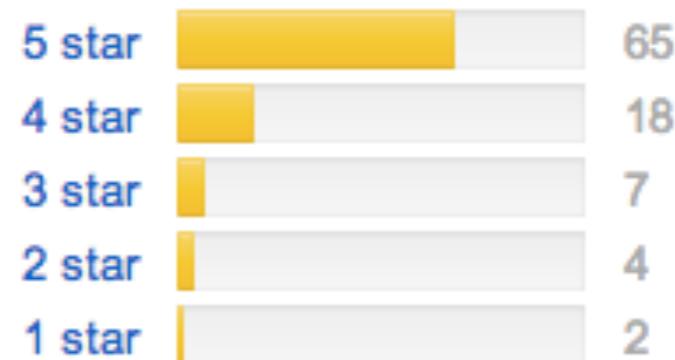
The Naked Prese...
Delivering Power
Present... by Gar
Reynolds
\$16.49



Customer Reviews

(96)

4.5 out of 5 stars



[See all 96 customer reviews ▾](#)

1 Zen
ions on
eynolds

People You May Know



Pixel Park

6 mutual friends

+1 Add Friend



Nolan Bushnell

8 mutual friends

+1 Add Friend



Juan Carlos Anorga

16 mutual friends

+1 Add Friend



Jiwoong Lee (Fragile)

31 mutual friends

+1 Add Friend



Ana Milat

+1 Add Friend



Kate Roberts

+1 Add Friend

See All

We created Pandora to put the [Music Genome Project](#) directly in your hands

**It's a new kind of radio –
stations that play only music you like**

Enter artist, genre or composer to create a station

Already have a Pandora account? [Sign in](#)

Company

[About](#)
[Advertising](#)
[Submit Your Music](#)
[Blog](#)
[Jobs](#)
[Investor](#)
[Press](#)
[Privacy](#)
[Your CA Privacy Rights](#)
[Terms](#)
[Ad Preferences](#)

Music Genre Stations

[Today's Hits](#)
[Today's Country](#)
[Today's Hip Hop And Pop Hits](#)
[Today's R&B And Hip Hop Hits](#)
[Contemporary Gospel](#)
[Classic Rock](#)
[Classical](#)
[Browse Genre Stations](#)

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Senior Data Analyst Job

Thomson Reuters - Bangalore, KA



Data Scientist/ Senior Data Scientist

HeadHonchos.com - Bangalore - IN



Adobe

Hiring Computer Scientist (Java) for...

Adobe - Noida



People You May Know



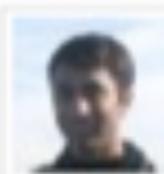
Rajeev Gupta, MBA Candidate '13 at Kelley School of Business

Connect



Alex Vauthey, Sr. Director of Engineering at LinkedIn

Connect



Bikram Bhatia, Graduate Research Assistant at University of Illinois at

Connect

[See more »](#)

[What to Watch](#)[My Subscriptions](#)[Music](#)

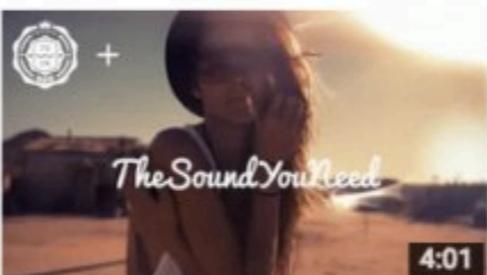
Recommended



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by The VGC
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281,611 views • 2 years ago

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Send Save Now Discard Draft autosaved at 6:01 PM (1 minute ago)

To: "Dopey" <hikingfan@gmail.com>, "Grumpy" <farnsworthpt@gmail.com>, "Doc" <surfingfan@gmail.com>

Also include: [Happy](#), [Sneezy](#), [Sleepy](#)

[Add Cc](#) | [Add Bcc](#)

Subject: New in Labs: Suggest more recipients

[Attach a file](#) [Add event invitation](#)

Market Basket



Market Basket Example



- Where should detergents be placed in the Store to maximize their sales?
- Are window cleaning products purchased when detergents and orange juice are bought together?
- Is soda typically purchased with bananas? Does the brand of soda make a difference?
- How are the demographics of the neighborhood affecting what customers are buying?

Image source: deepclimate.org

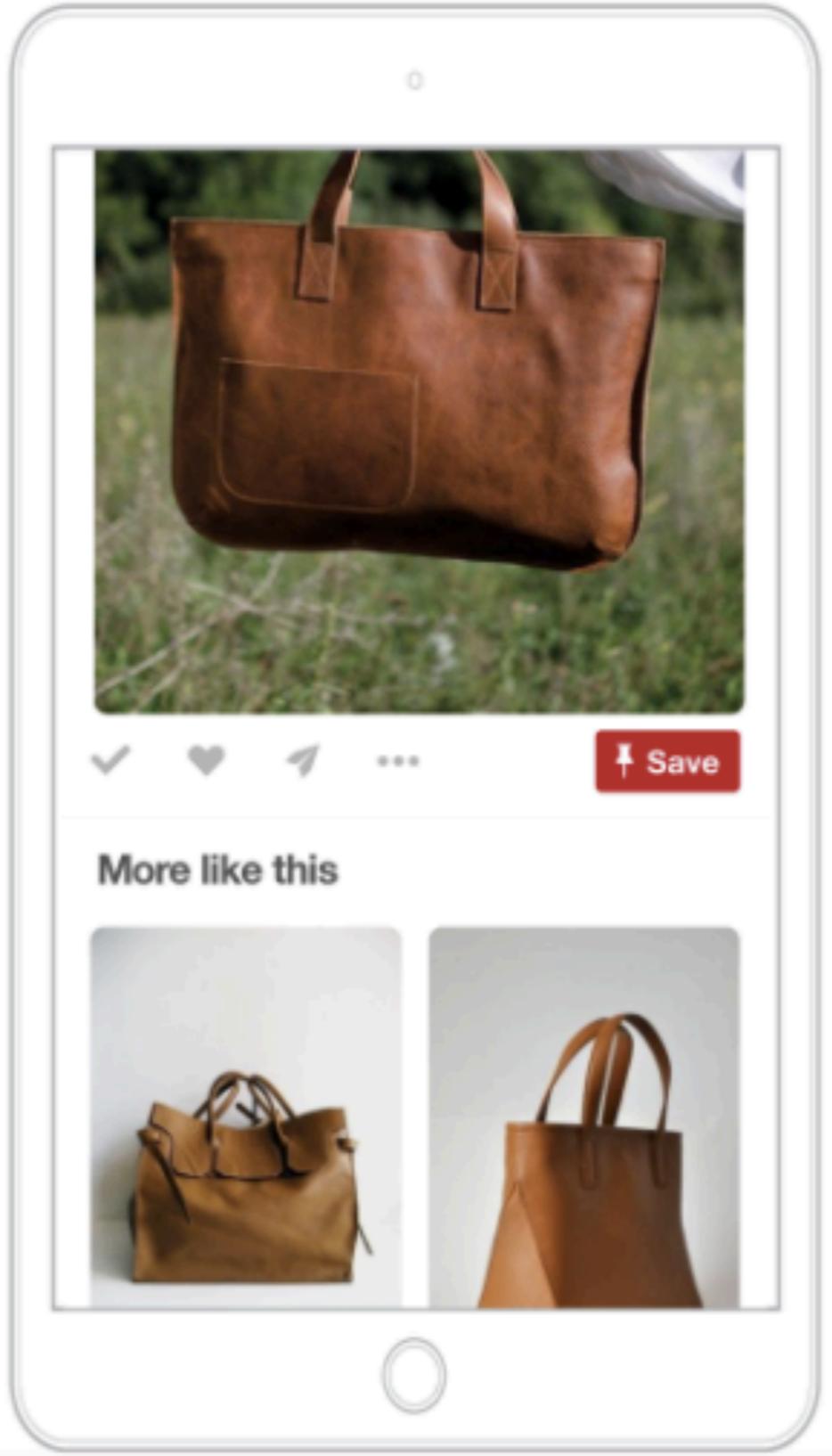


Figure 1. Related Pins recommendations for similar bags.





Zainab Mohiuddin, Blogger and Creator at Tinderella Worldwide (2017-present)

Answered Sep 22, 2017 · Author has 89 answers and 1m answer views

Hey Tinderos and Tinderellas! I'm a 26 year old woman who's been on tinder for 4 years and it's been **pretty awesome** for me. So awesome in fact, that I **research and blog about it** [↗](#).

How the tinder algorithm works

Tinder isn't like Instagram, where you just need to upload some bullshit images of your breakfast and buy some bots. Nope, the rules for your visibility are more **complex**.

- Tinder calculates how attractive you are by using an **ELO score**. What the hell is an ELO score? It's this rating system thing. Here, read [THIS](#) [↗](#).
- So it doesn't tell you how likely you are to be right-swiped, but **ranks you in terms of likability**. This is probably linked to it's **smart pictures** feature.
- The ELO score is made up of the % of people that like you as a whole, the % of matches you get and what the ELO scores of those people are.
- Your ELO score **determines who will be shown** your profile; so if you're a 7, you won't be shown to 4's but neither will you be shown to a 9. Harsh but true.
- You **can change your ELO** score by modifying your profile.
- You will **NEVER EVER** know your ELO score (because that's just too harsh).

Recommender Systems

- Two primary ways in which recommender systems may be formulated:
 1. **Prediction problem**
 2. **Ranking problem**, also referred as top-k recommendation problem

Types of Recommender Systems

Non-Personalized methods

Content-based methods

Collaborative-based methods

Knowledge-based methods

Non-Personalized methods

Conceptual goal:

Give me recommendation based on the users preferences
without using any information from the user itself.

Input:

Users ratings/Purchases/Behaviour

Content-based methods

Conceptual goal:

Give me recommendation based on the content (attributes) I have favored in my past ratings and actions

Input:

User ratings + item attributes

Collaborative-based methods

Conceptual goal:

Give me recommendation based on a collaborative approach that leverages the ratings and actions of my peers and myself

Input:

User ratings + community ratings

Knowledge-based methods

Conceptual goal:

Give me recommendation based on my explicit specification of the kind of content (attributes)

I want

Input:

User specifications + item attributes + domain knowledge

JAKE



Jake got a **beachball** & **sunglasses**

FUNMI



Funmi got a **polka-dot bikini**

NICK



Nick got a **beachball**...
What else might Nick like?

JAKE



Jake got a
beachball, sunglasses & flip flops

FUNMI



Funmi got a **bikini & flip flops**

NICK



Nick got a
beachball, sunglasses & flip flops

JOANNA



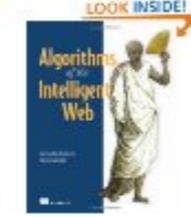
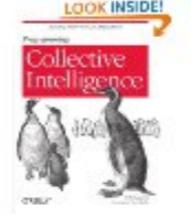
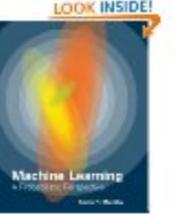
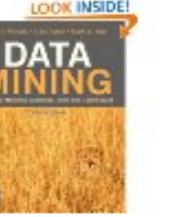
*What would you recommend
for Joanna?*

What is a good
Recommender system?

Recommender systems in e-commerce

- One Recommender Systems research question
 - **What should be in that list?**

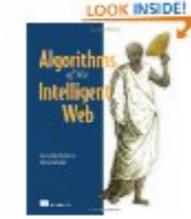
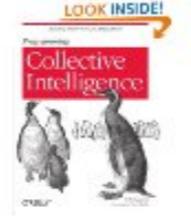
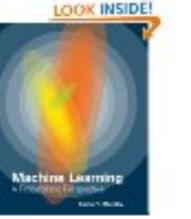
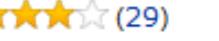
Customers Who Bought This Item Also Bought

 LOOK INSIDE!	 LOOK INSIDE!	 LOOK INSIDE!	 LOOK INSIDE!	 LOOK INSIDE!
Recommender Systems Handbook Francesco Ricci Hardcover \$167.73	Algorithms of the Intelligent Web Haralambos Marmanis Paperback \$26.76	Programming Collective Intelligence: ... Toby Segaran Paperback \$25.20	Machine Learning: A Probabilistic ... Kevin P. Murphy Hardcover \$81.00	Data Mining: Practical Machine Learning ... Ian H. Witten Paperback \$42.61

Recommender systems in e-commerce

- Another question both in research and practice
 - How do we know that these are **good recommendations?**

Customers Who Bought This Item Also Bought

 Recommender Systems Handbook Francesco Ricci Hardcover \$167.73	 Algorithms of the Intelligent Web Haralambos Marmanis  (14) Paperback \$26.76	 Programming Collective Intelligence: ... Toby Segaran  (91) Paperback \$25.20	 Machine Learning: A Probabilistic ... Kevin P. Murphy  (15) Hardcover \$81.00	 Data Mining: Practical Machine Learning ... Ian H. Witten  (29) Paperback \$42.61
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Recommender systems in e-commerce

- This might lead to ...
 - What is a **good** recommendation?
 - What is a **good** recommendation **strategy**?
 - What is a **good** recommendation **strategy for my business**?

Customers Who Bought This Item Also Bought



Goals of the recommender systems?

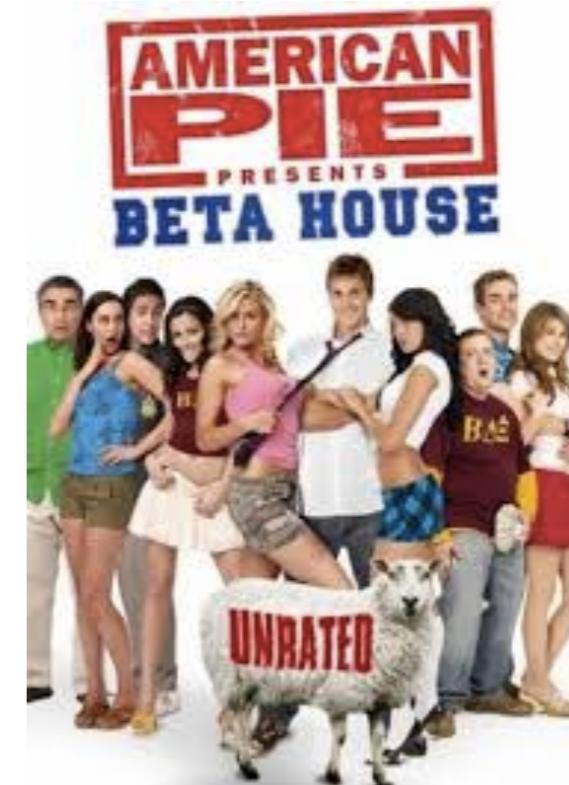
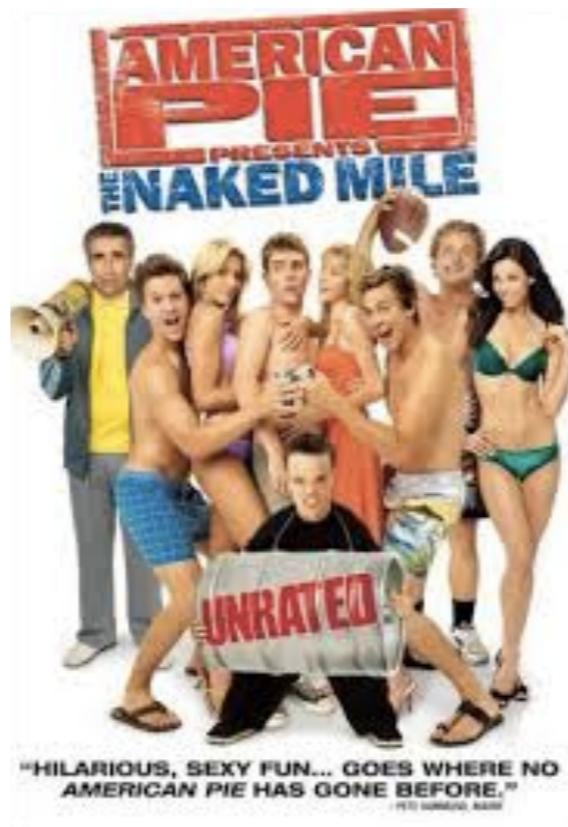
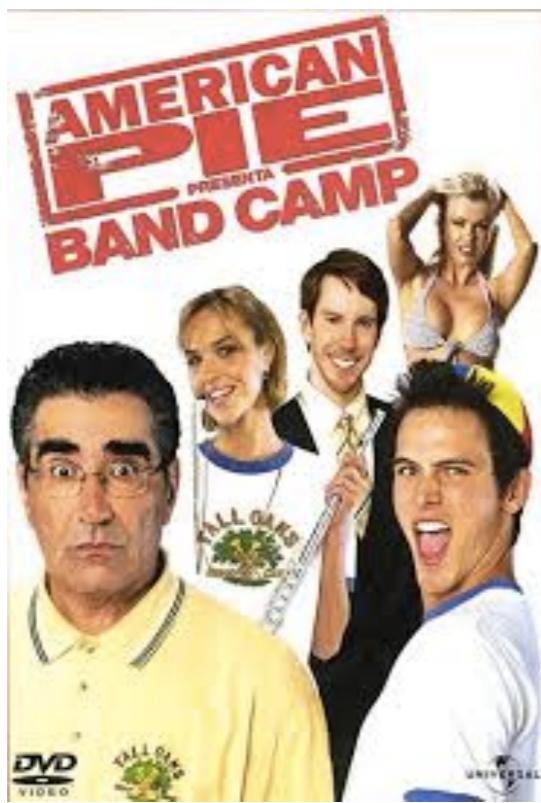
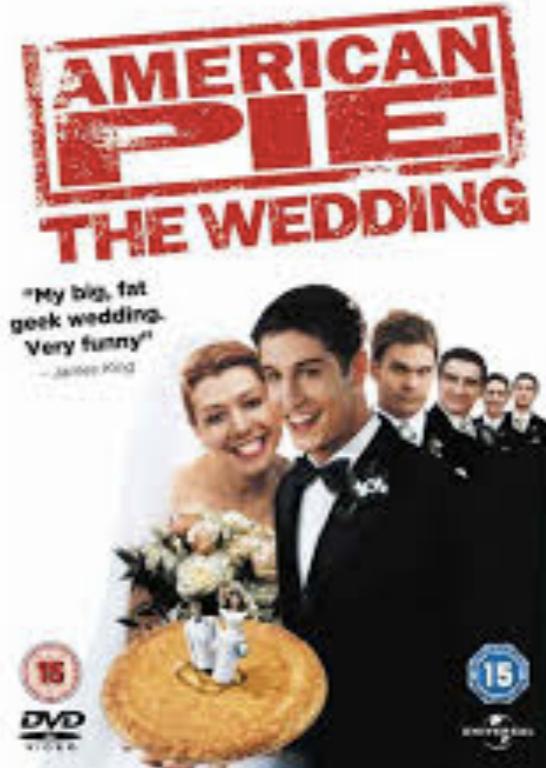
What do you think?

Goals of the recommender systems?

1st: Increase product sales or revenue

ok,
but this is the business-centric goal.

How can we evaluate a recommender?



How can we **evaluate** a recommender system?

The common operational and technical goals are:

Relevance

Novelty

Serendipity (pleasant and unexpected)

Recomendation diversity

Which are the measures in the practice?

- Accuracy
- Total number of sales
- Promotion of certain items
- Customer return rates
- Customer satisfaction and loyalty

CAUTION

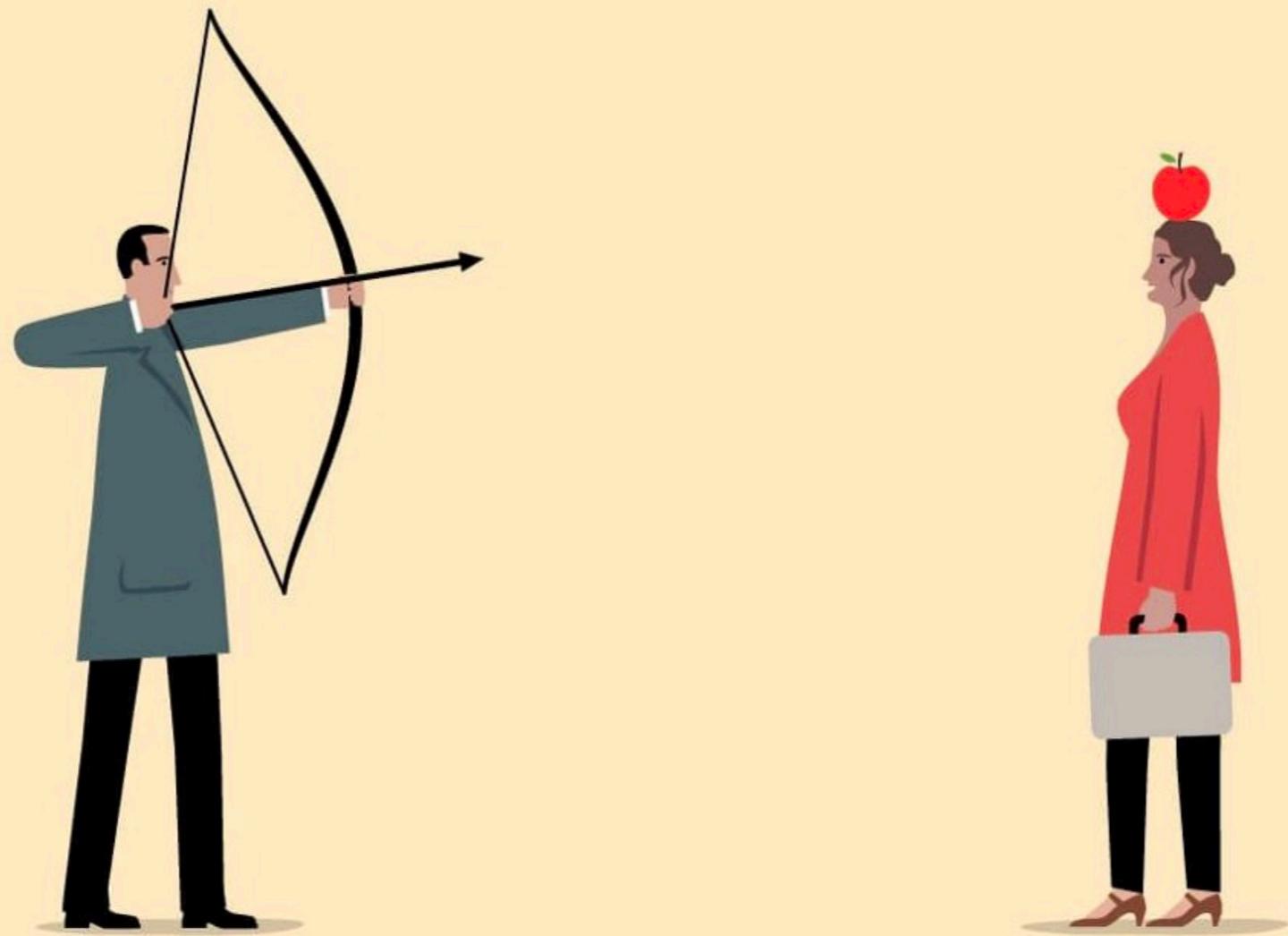
**RecSys can
change
user behaviors**

Evaluating Recommender systems

- This is one of the **most critical** step when building a recommender system.
- A single criterion cannot capture many goals of the designer.
- ***Online*** and ***offline*** methods are used

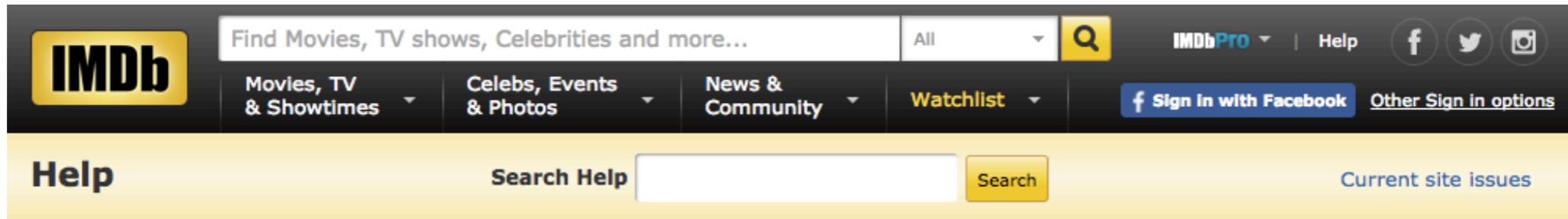
One critical point!!

Trust is critical!!



Are you recommending me
what is **good for me** or what is **good for you**?

Be transparent!

A screenshot of the IMDb website's help section. At the top, there is a navigation bar with the IMDb logo, a search bar, and links for "IMDbPro", "Help", and social media icons (Facebook, Twitter, Instagram). Below the navigation bar, the page title "Help" is displayed, along with a search bar and a "Search" button. A link to "Current site issues" is also present.

Personalized Recommendations Frequently Asked Questions

IMDb makes personalized recommendations to help you discover movies and TV shows that you will love.

Where can I find my Watchlist and my ratings?

Find your [Watchlist](#) using the *Watchlist* link in the top right of the menu at the top of every page on the IMDb site. Find [your ratings](#) by using the *Your Ratings* item when you hover over your account name in the menu bar.

How does IMDb choose personalized recommendations?

First, we take all of the movies and TV shows that you have either [rated](#) or added to your [Watchlist](#). Then, we compare your data to ratings made by other users. We can then find movies and TV shows that people with similar tastes to you like. For each recommendation, you can see a list of the movies or TV shows upon which the recommendation was based. You have either rated these titles highly, or added them to your Watchlist.

How does IMDb know what I "showed interest in"?

When you give a movie a positive rating or add a movie to your Watchlist, we track that as a movie that you are interested in.

Why do I see items in my recommendations that I've already rated?

If we do not have any recommendations for you (either because you have logged out or because you have not rated many items), IMDb will show you a list of items that many people have seen so that you can enter ratings to get recommendations. When this happens, the widget will be titled *Rate and Get Recommendations* instead of *Recommended for You*.

Where can I find personalized recommendations?

You can see your personalized recommendations at the bottom of your [Watchlist](#) or the bottom of your [Ratings History](#). If you are logged in they are also visible on the IMDb home page in the section *Recommended for You*.

You are recommending a movie that I really do not want to see. How can I make it go away?

Click on the 'no' button.

How can I improve my personalized recommendations?

To improve your personalized recommendations, [find and rate the movies that you love](#).

I've just rated a whole bunch of titles. How long will it take for those ratings to impact my recommendations?

New ratings will have an immediate impact on your recommendations. After rating and adding titles to your Watchlist, reload the page to update your recommendations.

How do I suggest features or changes to personalized recommendations?

Please give us feedback via the [contact form](#).

Be transparent!



Note:

- If you're unsure why we're recommending a particular item, go to [Your Amazon.com](#), then click **Why Recommended?** beneath the item. We'll tell you which of your previous ratings, likes or purchases influenced our suggestion so you can revise your ratings or exclude those purchases.
- You can tell us about items you like, even if you didn't purchase it on Amazon.com. Locate the item on [Amazon.com](#), then write a review and add a rating. We'll use that rating to help provide you better recommendations.

Be transparent!

10.



Apple MC747Z/A - Adaptador de corriente para MacBook y 13" MacBook Pro (45W), blanco

de Apple (21 de octubre de 2010)

Valoración media de los clientes: ★★★★☆ (13)

En stock

Precio: EUR 79,00

[22 de 2^a mano y nuevo\(s\) desde EUR 68,52](#)

Añadir a la cesta

Añadir a la Lista de deseos

Ya lo tengo No me interesa ★★★★★ Valorar este producto

Recomendado porque has añadido Apple MD592Z/A - Adaptador de corriente para MacBook Air a tu cesta ([Cambiar esto](#))

11.



Caja de 20 paquetes de toallitas Zig Zag secamanos doble capa

de Clim Profesional (13 de noviembre de 2014)

Valoración media de los clientes: ★★★★☆ (3)

En stock

Precio: EUR 25,90

Ofrecido por [CLIM Profesional productos de limpieza](#)

Añadir a la cesta

Añadir a la Lista de deseos

Ya lo tengo No me interesa ★★★★★ Valorar este producto

Recomendado porque has añadido Apple MD592Z/A - Adaptador de corriente para MacBook Air a tu cesta ([Cambiar esto](#))

12.



DELL Vostro 3559 2.3GHz i5-6200U - Ordenador portátil de 15.6" (i5-6200U, 1366 x 768 Pixeles, teclado Español QWERTY), color Negro

de Dell (10 de marzo de 2016)

Valoración media de los clientes: ★★★★☆ (1)

En stock

Precio: EUR 566,00

Ofrecido por [CanalPC Informática](#)

Añadir a la cesta

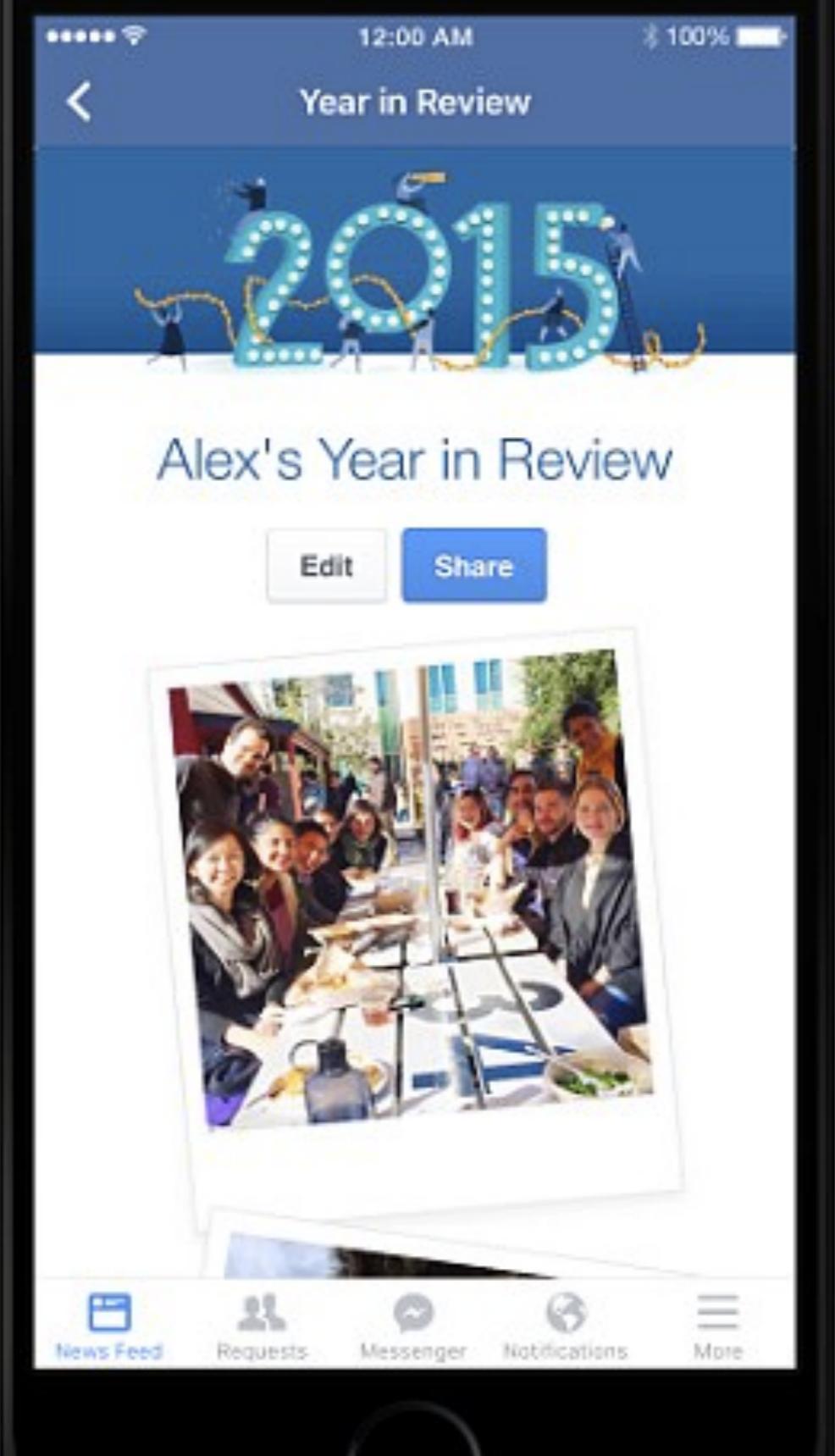
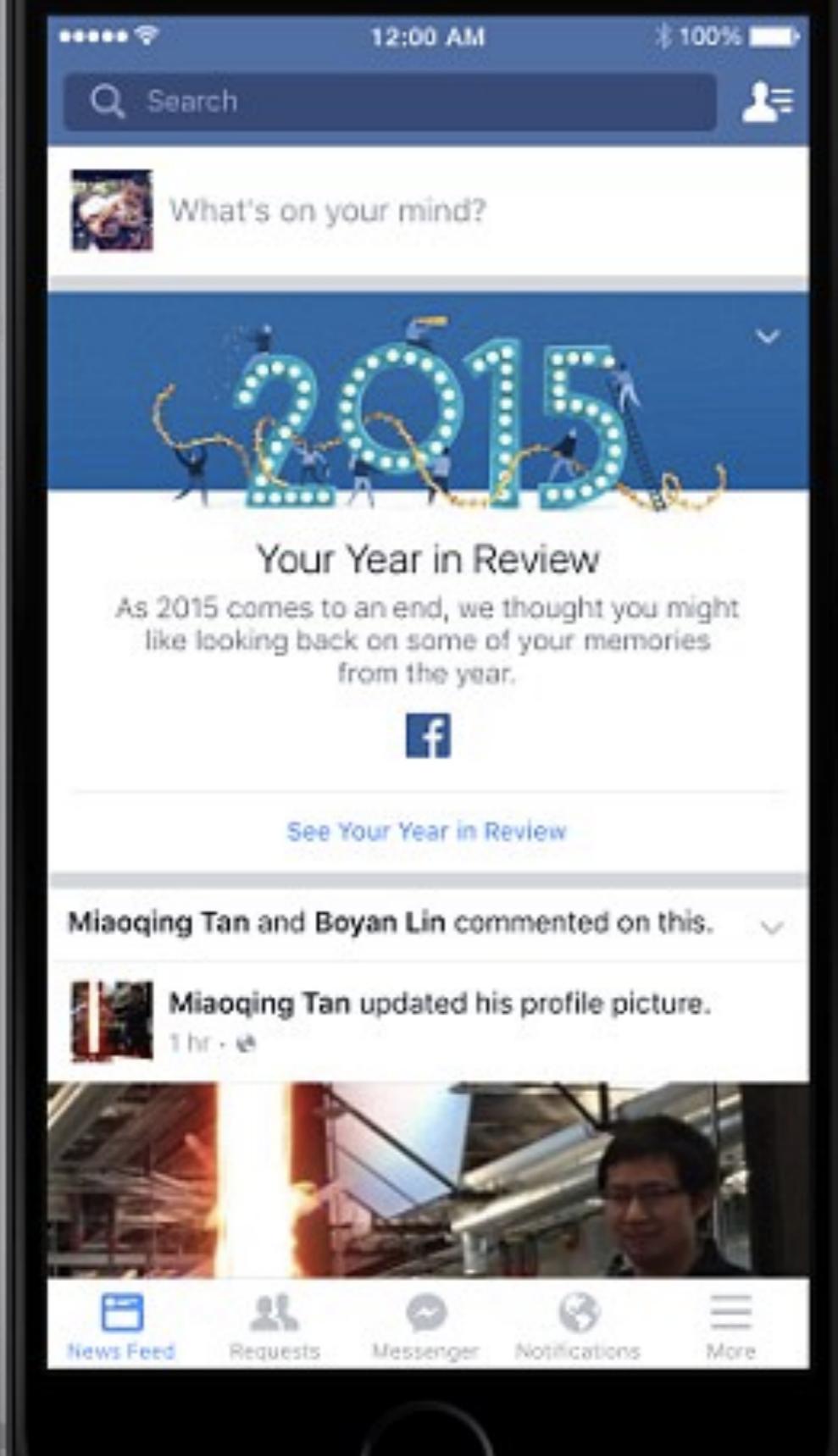
Añadir a la Lista de deseos

Ya lo tengo No me interesa ★★★★★ Valorar este producto

Recomendado porque has añadido Apple MD592Z/A - Adaptador de corriente para MacBook Air a tu cesta ([Cambiar esto](#))

Another critical point!!

Not everything is allowed!!



Facebook Apologizes For Its 'Year In Review' Approach



Amit Chowdhry, CONTRIBUTOR

[FULL BIO ▾](#)

Opinions expressed by Forbes Contributors are their own.

This month, [Facebook](#) launched a customizable slideshow feature called *Year In Review*. To increase the engagement in *Year In Review*, Facebook automatically created a slideshow with suggested photos for each of their users. Facebook's algorithms decided the suggested photos based on the month each photo was uploaded and the number of "likes" they received. The default tagline for the *Year In Review* slideshow is "It's been a great year! Thanks for being a part of it." Unfortunately, many Facebook users had a tumultuous year instead of a "great" year and *Year In Review* served as a reminder of traumatic circumstances.

This year, web designer Eric Meyer lost his six-year-old daughter Rebecca to brain cancer so he did not have any interest in creating a *Year In Review* slideshow to share with his Facebook friends. However, Facebook's algorithms automatically created a slideshow for him with his daughter's face in the center of the first slide with some holiday-themed clip art. Facebook displayed the slideshow at the top of his News Feed on December 24th, asking him if he wanted to customize and share his *Year In Review*.

"I know, of course, that this is not a deliberate assault. This inadvertent algorithmic cruelty is the result of code that works in the overwhelming majority of cases, reminding people of the awesomeness of their years, showing them selfies at a party or whale spouts from sailing boats or the marina outside their vacation house," said Meyer in a [blog post](#). "But for those of us who lived through the death of loved ones, or spent extended time in the hospital, or were hit by divorce or losing a job or any one of a hundred crises, we might not want another look at this past year."

Which data recommender systems use?

Two types of data

Explicit



Filling out a form



Making a purchase



Favoriting an item

Implicit



Browsing activity



Searches



Location

Basic notation

- Some basic notation we will use throughout the course:
 - $m \times n$ rating matrix denoted by R , where m is the total number of users and n is the total number of items.
 - The (i, j) entry of R indicates the rating of user i to the item j , and it is denoted by r_{ij}
 - When the entry (i, j) , is estimated, or predicted, it is denoted by \hat{r}_{ij} with a “hat” symbol indicating that it is estimated.

Innovation on Recommender Systems





ACM RecSys 2018

The 12th ACM Recommender Systems Conference will take place in Vancouver, Canada from Oct 2-7, 2018.

LATEST NEWS

Jan 12, 2018: Apply for a [SIGCHI student travel grant](#) by Feb 01, 2018 to receive travel support for RecSys 2018!

Dec 12, 2017: The call for [doctoral symposium submissions](#) is now online!

Nov 09, 2017: Calls for [long and short papers](#), [workshop proposals](#) and [demos](#) are now online!

Aug 30, 2017: See you all next year in Vancouver, Canada at [RecSys 2018](#) from Oct 2-7, 2018!

SHORTCUTS TO CONFERENCES

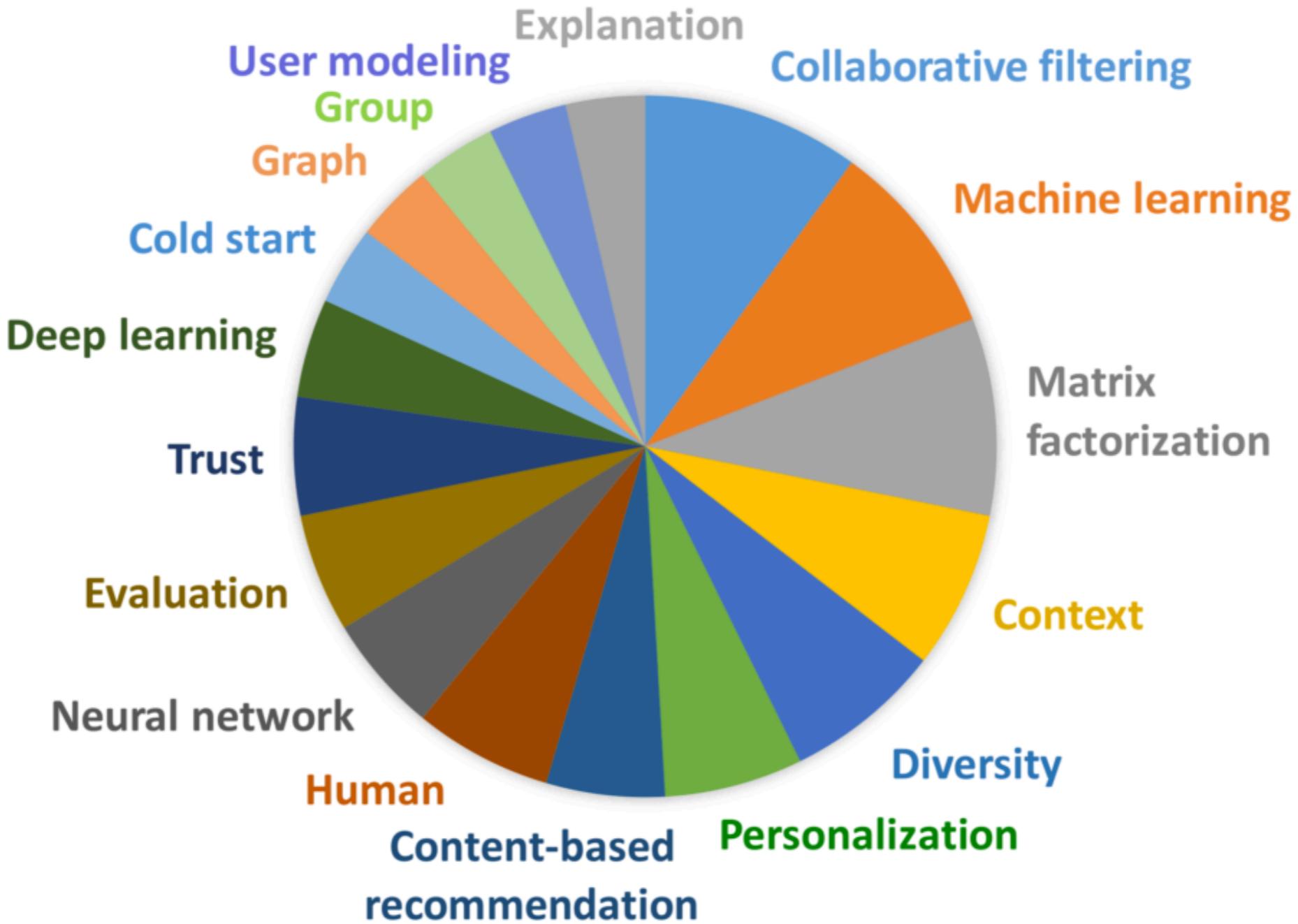
- [RecSys 2018](#) (Vancouver)
- [RecSys 2017](#) (Como)
- [RecSys 2016](#) (Boston)
- [RecSys 2015](#) (Vienna)
- [RecSys 2014](#) (Silicon Valley)
- [RecSys 2013](#) (Hong Kong)



About

The RecSys Challenge 2019 will be organized by [trivago](#), [TU Wien](#), [Politecnico di Milano](#), and [Karlsruhe Institute of Technology](#). trivago is a global hotel search platform focused on reshaping the way travelers search for and compare hotels, while enabling advertisers of hotels to grow their businesses by providing access to a broad audience of travelers via our websites and apps. trivago has established 55 localized platforms in over 190 countries and provides access to over two million hotels, including alternative accommodations, with prices and availability from over 400+ booking sites and hotel chains.

This year's challenge focuses on travel metasearch. The goal of this challenge is to develop a session-based and context-aware recommender system using various input data to provide a list of accommodations that will match the needs of the user.



Accepted Papers - Top Author Keywords



\$60,000 • 1,717 teams

Santander Product Recommendation

Wed 26 Oct 2016

Merger and Entry Deadline

Wed 21 Dec 2016 (6.5 days to go)

Dashboard

- Home
- Data
- Make a submission

Information

- Description
- Evaluation
- Rules
- Prizes
- Timeline

Forum

Kernels

- New Script
- New Notebook

Leaderboard

[Competition Details](#) » [Get the Data](#) » [Make a submission](#)

Can you pair products with people?

Ready to make a downpayment on your first house? Or looking to leverage the equity in the home you have? To support needs for a range of financial decisions, [Santander Bank](#) offers a lending hand to their customers through personalized product recommendations.



Public Leaderboard

1. Tom Van de Wiele
2. In Public Leaderboard We Trust
3. yoniko
4. idle_speculation
5. Alejo y Miro
6. Jared Turkewitz and BreakfastPirate
7. colun
8. Jack (Japan)
9. Sameh Faidi
10. DL aware

Under their current system, a small number of Santander's customers receive many recommendations while many others rarely see any resulting in an uneven customer experience. In their second competition, Santander is challenging Kagglers to predict which products their existing customers will use in the next month based on their past behavior and that of similar customers.

With a more effective recommendation system in place, Santander can better meet the individual needs of all customers and ensure their satisfaction no matter where they are in life.

Disclaimer: This data set does not include any real Santander Spain's customer, and



Expedia Hotel Recommendations

Which hotel type will an Expedia customer book?

\$25,000 · 1,974 teams · 8 months ago

[Overview](#)[Data](#)[Kernels](#)[Discussion](#)[Leaderboard](#)[More](#)[Submit Predictions](#)

Overview

[Description](#)[Evaluation](#)[Prizes](#)[Timeline](#)

Planning your dream vacation, or even a weekend escape, can be an overwhelming affair. With hundreds, even thousands, of hotels to choose from at every destination, it's difficult to know which will suit your personal preferences. Should you go with an old standby with those pillow mints you like, or risk a new hotel with a trendy pool bar?



Expedia wants to take the proverbial rabbit hole out of hotel search by providing personalized hotel recommendations to their users. This is no small task for a site with hundreds of millions of visitors every month!

Currently, Expedia uses search parameters to adjust their hotel recommendations, but there aren't enough customer specific data to personalize them for each user. In this competition, Expedia is challenging Kagglers to contextualize customer data and predict the likelihood a user will stay at 100 different hotel groups.

The data in this competition is a random selection from Expedia and is not representative of the overall statistics.

Complex Task & Research

Cross-Domain Recommendation

Context-aware Recommendation

Constraint-Based Recommendation

Recommendation and Privacy

Group Recommendation

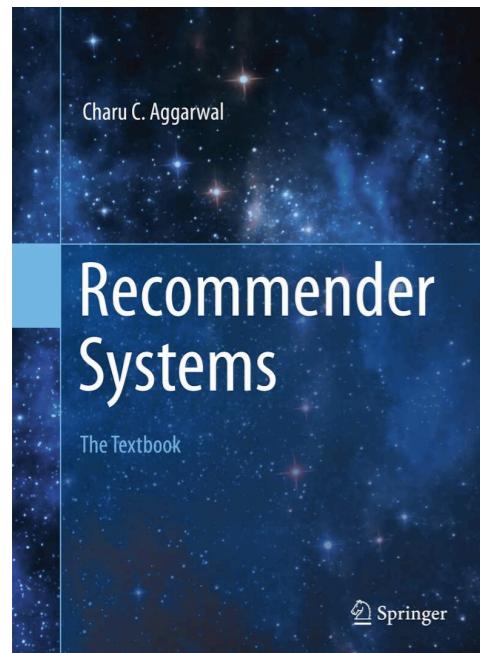
Recommendation as Intelligent Task Routing

Impact of Recommendations

New Domains for Recommendation

Recommendation and Social Networks

Bibliography



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Google