The Long Tail and Recommender Systems

# Hits vs. Niche

- Giant retailers of books, music, etc.
- <u>Question</u>: are most sales being generated
  - by a small set of items that are enormously popular ("hits", blockbusters), or
  - by a much larger population of items that are each individually less popular ("niche products" appealing to a small segment of the audience)?
- <u>Answer</u>:
  - Despite stereotype of media business: only blockbusters matter
  - We observe that the total sales volume of *unpopular items, taken together, is* very significant

# The Long Tail

- Chris Anderson (2004)
  - Internet-based distribution is driving the media and entertainment industries to a "long tail" of obscure products driving the bulk of sales
- Amazon or Netflix sell an astronomical diversity of products (no restrictions imposed by physical stores) even when very few of them generate much volume on their own
- Quantifying the importance of the Long Tail comes down to an analysis of power laws

# Visualizing the Long Tail

- Our original definition of the popularity curve:
   As a function of k, what fraction of items have popularity exactly k?
- Modify our original definition slightly:

As a function of k, what number of items have popularity at least k?



• If the original function was a power-law, then this new one is too (we showed that in previous lecture)



- Question we ask: As you look at larger and larger sales volumes, how few books do you find?
- Questions we want to ask: As you look at less and less popular items, what sales volumes do you see?



- Interchange the roles of the x- and y-axes
- New curve says: The j-th most popular book has sold k copies

# Visualizing the Long Tail



- Order by "sales rank" and look at the popularity as we move out to larger and larger sales ranks -> into the niche products
- The characteristic shape gives the name "Long Tale"
- Significantly more area under the right (niche products) compared to the left part (hits)

# Need for Recommendations

- Companies make money from a giant inventory of niche products when customers are aware of these products and have some reasonable way to explore them
- Recommender systems that companies like Amazon and Netflix have popularized can be seen as integral to their business strategies
  - They expose people to items that may not be generally popular, but which match user interests as inferred from their history of past purchases

# **Recommender Systems**

Need For Recommenders

 Rapid Growth of Information
 Lots of Options for Users

- Input Data
  - A set of users  $U = \{u_1, ..., u_N\}$
  - A set of items I= $\{i_1, ..., i_M\}$
  - The rating matrix  $R = [r_{u,i}]_{NxM}$



# Problem Definitions in RSs

Predicting the rating on a target item for a given user (i.e. Predicting John's rating on Star Wars Movie).



Recommending a List of items to a given user

(i.e. Recommending a list of movies to John for watching).



### What book should I buy?

#### **Customers Who Bought This Item Also Bought**





Reckoning with Risk: Learning to Live with Unce... by Gerd Gigerenzer Gut Feelings: The Intelligence of the Unconscious by Gerd Gigerenzer £10.27 Bounded Rationality: The Adaptive Toolbox (Dahl... b G Gigerenzer £20.95

#### What Do Customers Ultimately Buy After Viewing This Item?



4

#### 68% buy Simple Heuristics That Make Us Smart (Evolution & Cognition) £18.99

20

Gut Feelings: Short Cuts to Better Decision Making

17% buy Gut Feelin £6.74

#### 9% buy

Influence: The Psychology of Persuasion



## What movie should I watch?



- The Internet Movie Database (IMDb) provides information about actors, films, television shows, television stars, video games and production crew personnel.
- Owned by Amazon.com since 1998
- 796,328 titles and 2,127,371 people
- More than 50M users per month.

### The Nextflix prize story

- In October 2006, Netflix announced it would give a \$1 million to whoever created a movie-recommending algorithm 10% better than its own.
- Within two weeks, the DVD rental company had received 169 submissions, including three that were slightly superior to Cinematch, Netflix's recommendation software
- After a month, more than a thousand programs had been entered, and the top scorers were almost halfway to the goal
- But what started out looking simple suddenly got hard. The rate of improvement began to slow. The same three or four teams clogged the top of the leader-board.
- Progress was almost imperceptible, and people began to say a 10 percent improvement might not be possible.
- Three years later, on 21<sup>st</sup> of September 2009, Netflix announced the winner.



## What news should I read?

Yahoo!	My Yaho	o!   Mail   More 🔻 🛛 Make Y! your home page						Welcome, fmr59 Sign Out   Help				
YAHOO! NEWS							Search: [				Web Se	arch
Home	U.S.	Busines	s Wo	rld Enter	tainmen	t Spor	ts Tech	Politics	Election	s Science	Health	
Most F	Popular											
Video	Photos	Opinion	Local	Odd News	Comics	Travel	Weather	People of th	ne Web 🗅	You Witness N	lews	
Site Ind	ex											
		Search:				All News		Search Advanced				
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### Black boxes found in Thai plane crash

AP - 59 minutes ago

PHUKET, Thailand - Authorities on Monday found the two flight data recorders from a plane that crashed in stormy weather on the resort island of Phuket, killing 90 people, including 54 foreign tourists.

 sLIDESHOW: Thai plane crashes in Phuket
 vIDEO: First Person: 'I knew the plane was in trouble' AP



### Bush to pick Mukasey as attorney general

AP - 42 minutes ago

WASHINGTON - Michael Mukasey, President Bush's pick to replace Alberto Gonzales as attorney general, is not expected to prompt the confirmation battle that Senate Democrats



Reuters

Enlarge Photo 🕀

### `Sopranos' wins best drama Emmy

#### AP - Mon Sep 17, 12:41 AM ET

LOS ANGELES - "The Sopranos" turned its startling cut-to-black final season into Emmy gold Sunday, winning the best drama series award, and newcomer "30 Rock" was named best comedy series.



0.J. Simpson ordered held without bail

### Associated Press » All Video

- Emmy Red Carpet: What celebs are watching
- Global action for Darfur 👘 🖸
- Greenspan Slams Bush in D
   New Book

Caught on Tape: Drag racer cheats death

### Where should I spend my vacation?







### Tripadvisor.com

I would like to escape from this ugly an tedious work life and relax for two weeks in a sunny place. I am fed up with these crowded and noisy places ... just the sand and the sea ... and some "adventure".

I would like to bring my wife and my children on a holiday ... it should not be to expensive. I prefer mountainous places... not too far from home. Children parks, easy paths and good cuisine are a must.

I want to experience the contact with a completely different culture. I would like to be fascinated by the people and learn to look at my life in a totally different way.

## **Collaborative Filtering**



# **Collaborative Filtering**

	Star Wars	Hoop Dreams	Contact	Titanic
Joe	5	2	\$	4
John	2	5		3
Al	2	2	4	2
Nathan	5	1	5	į

The problem of collaborative filtering is to predict how well a user will like an item that he has not rated given a set of historical preference judgments for a community of users

# User-based CF

Calculate the similarity (weight) w<sub>u,v</sub> between the active user u and all other users v:

$$w_{u,v} = \frac{\sum_{i \in I} (r_{u,i} - \overline{r}_u) (r_{v,i} - \overline{r}_v)}{\sqrt{\sum_{i \in I} (r_{u,i} - \overline{r}_u)^2} \sqrt{\sum_{i \in I} (r_{v,i} - \overline{r}_v)^2}}$$

- The *i EI* summations are over the items that both the users *u* and *v* have rated
- Generate a top-N recommendation:
  - Find k most similar users (nearest neighbors)
  - Identify a set of items, C, purchased by the nearest neighbors, together with their frequency
  - Recommend the top-N most frequent items in C that the active user u has not already purchased

# **Evaluation metrics**

- Precision of top-N item recommendation:
  - For the active (test) user assume a subset of its ratings as known and the rest as unknown
  - Generate top-N item recommendation based on the known ratings
  - Find the number X of top-N recommender items that also belong in the unknown items
  - Precision is X/N