1-1: Introduction to Recommender Systems

Today's Learning Objectives

- Understand what a recommender system is
- Some history and background
- NOT: course details (next lecture)

A Bit of History

- Ants, Cavemen, and Early Recommender Systems
 - The emergence of critics
- Information Retrieval and Filtering
- Manual Collaborative Filtering
- Automated Collaborative Filtering
- The Commercial Era

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Information Retrieval

- Static content base
 - Invest time in indexing content
- Dynamic information need
 - Queries presented in "real time"
- Common approach: TFIDF
 - Rank documents by term overlap
 - Rank terms by frequency

Information Filtering

- Reverse assumptions from IR
 - Static information need
 - Dynamic content base
- Invest effort in modeling user need
 - Hand-created "profile"
 - Machine learned profile
 - Feedback/updates
- Pass new content through filters

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Collaborative Filtering

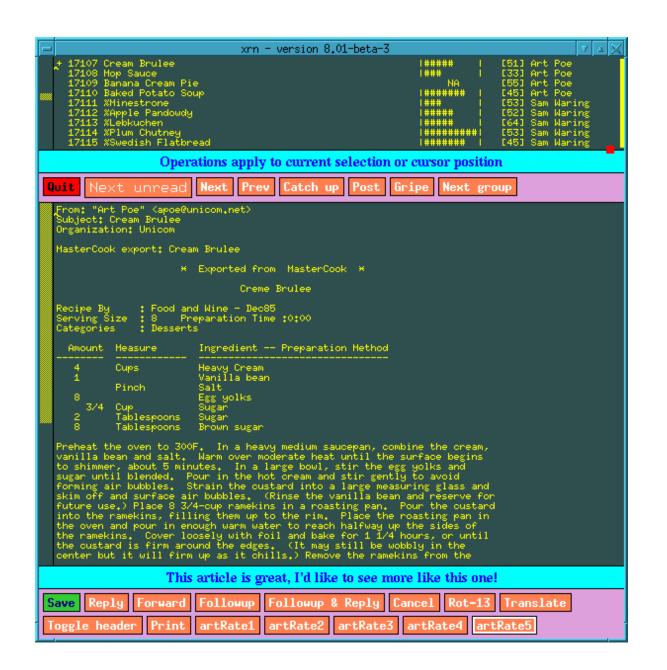
- Premise
 - Information needs more complex than keywords or topics: quality and taste
- Small Community: Manual
 - Tapestry database of content & comments
 - Active CF easy mechanisms for forwarding content to relevant readers

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Automated CF

- The GroupLens Project (CSCW '94)
 - ACF for Usenet News
 - users rate items
 - users are correlated with other users
 - personal predictions for unrated items
 - Nearest-Neighbor Approach
 - find people with history of agreement
 - assume stable tastes



Does it Work?

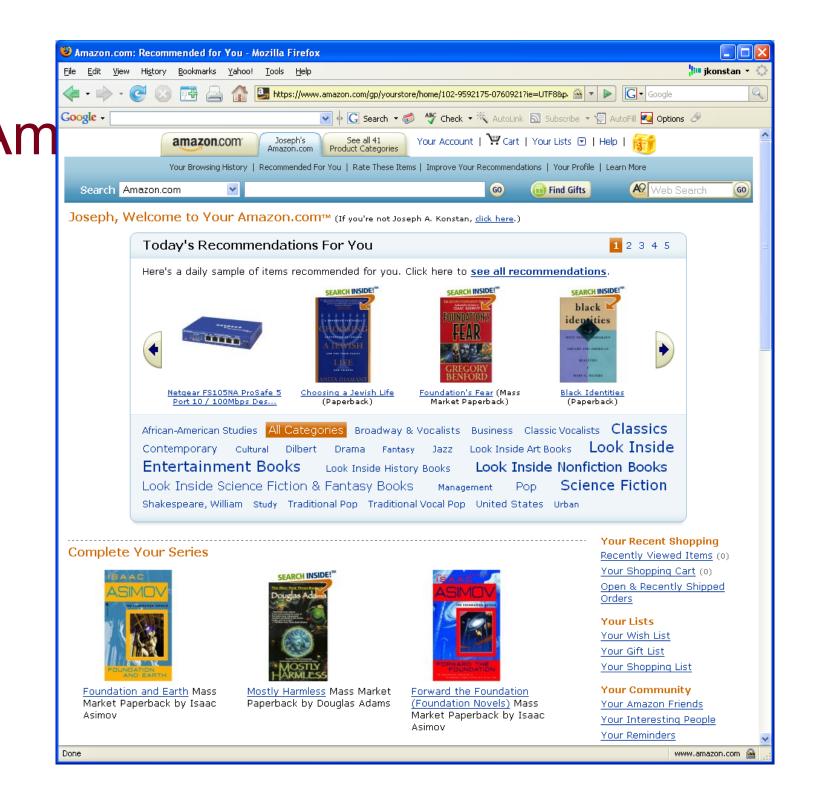
- Yes: The numbers don't lie!
 - Usenet trial: rating/prediction correlation
 - rec.humor: 0.62 (personalized) vs. 0.49 (avg.)
 - comp.os.linux.system: 0.55 (pers.) vs. 0.41 (avg.)
 - rec.food.recipes: 0.33 (pers.) vs. 0.05 (avg.)
 - Significantly more accurate than predicting average or modal rating.
 - Higher accuracy when partitioned by newsgroup

It Works Meaningfully Well!

- Relationship with User Behavior
 - Twice as likely to read 4/5 than 1/2/3
- Users Like GroupLens
 - Some users stayed 12 months after the trial!

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A TOUR OF MOVIELENS (CLASSIC COLLABORATIVE FILTERING)

K-Nearest Neighbor User-User



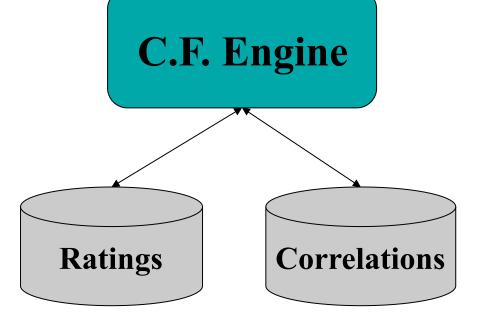






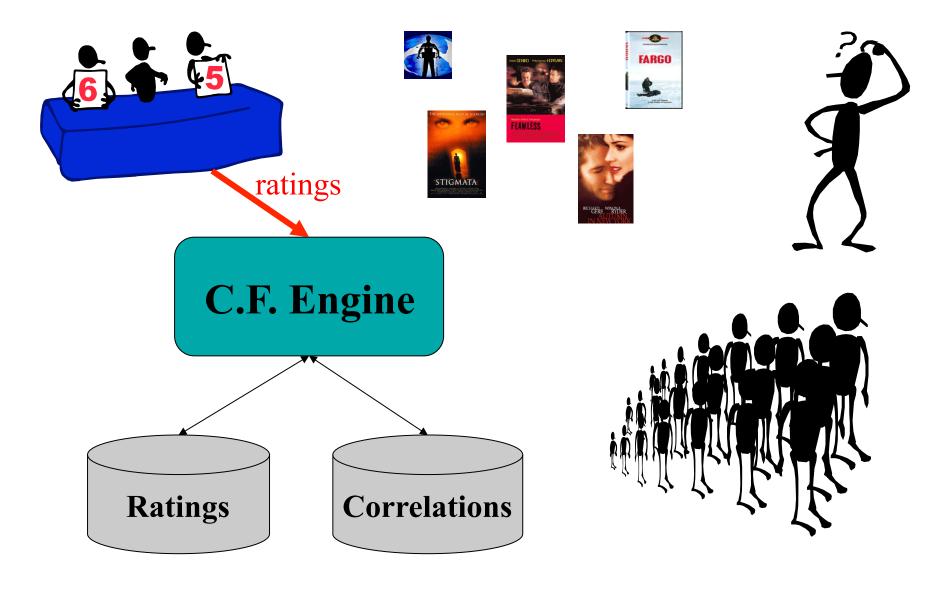




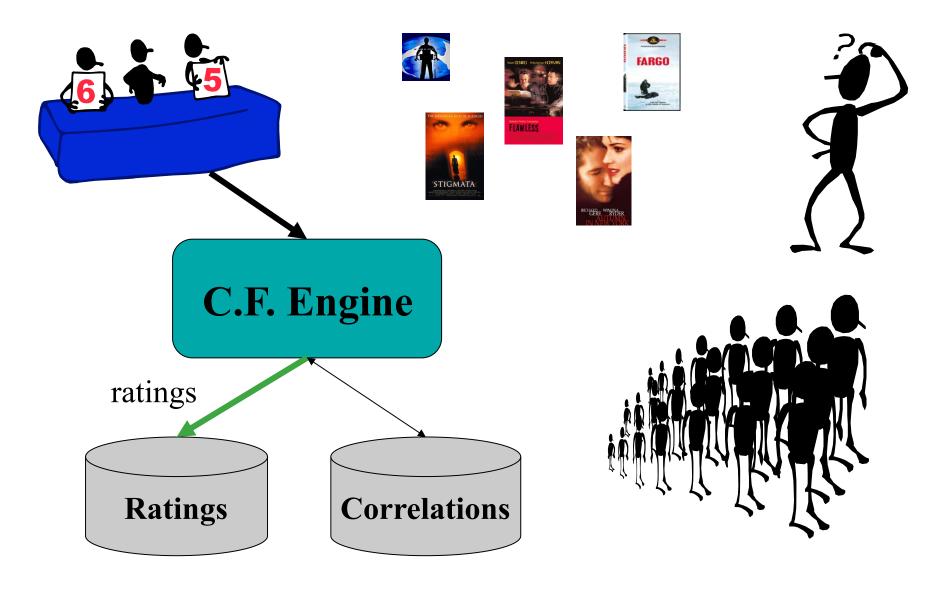




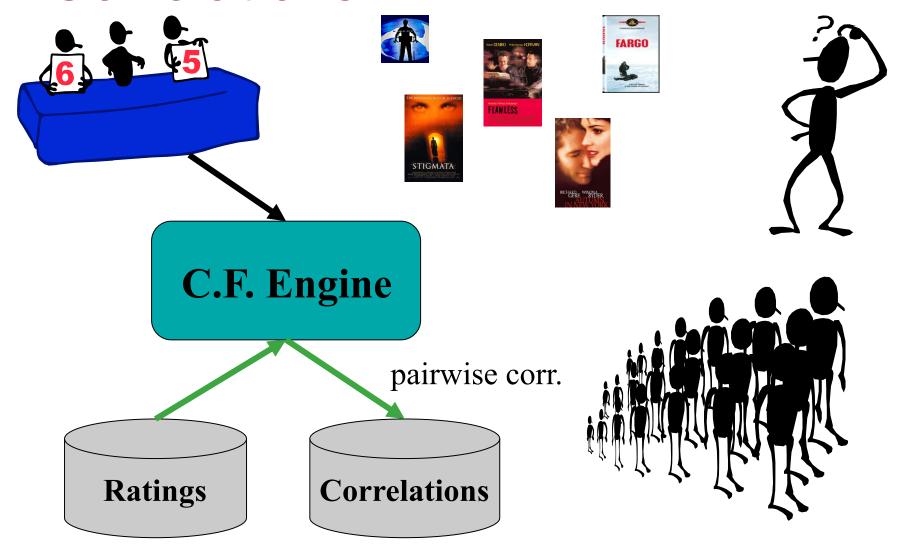
CF Classic: Submit Ratings



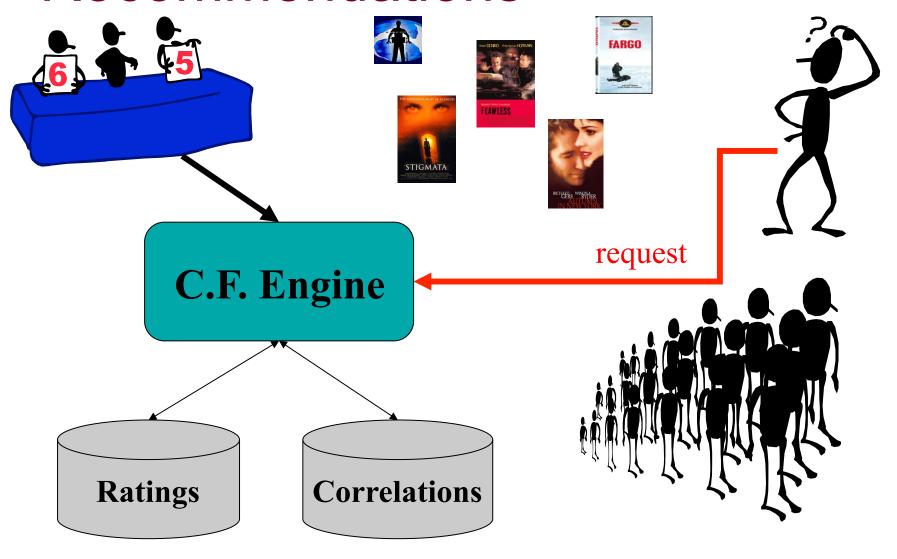
CF Classic: Store Ratings



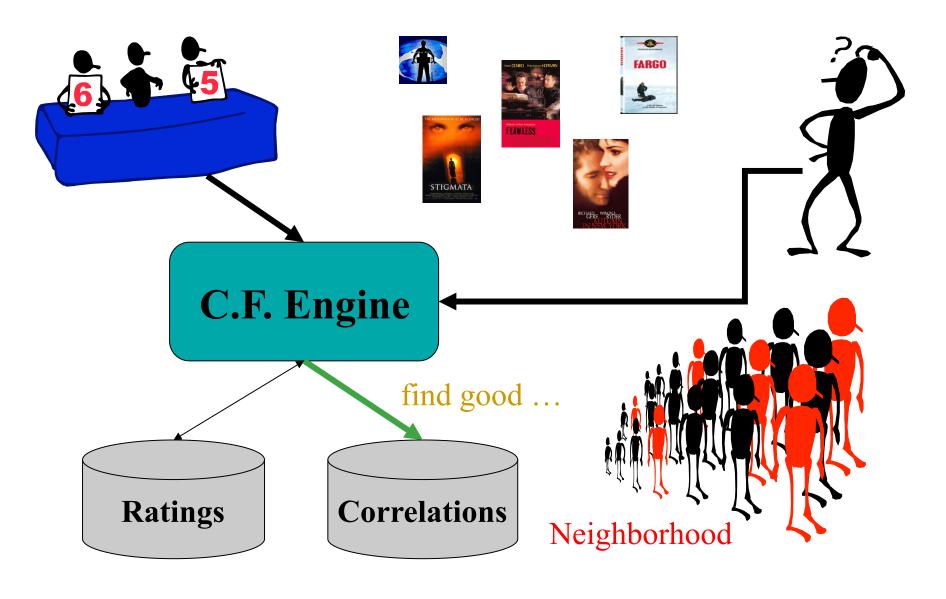
CF Classic: Compute Correlations



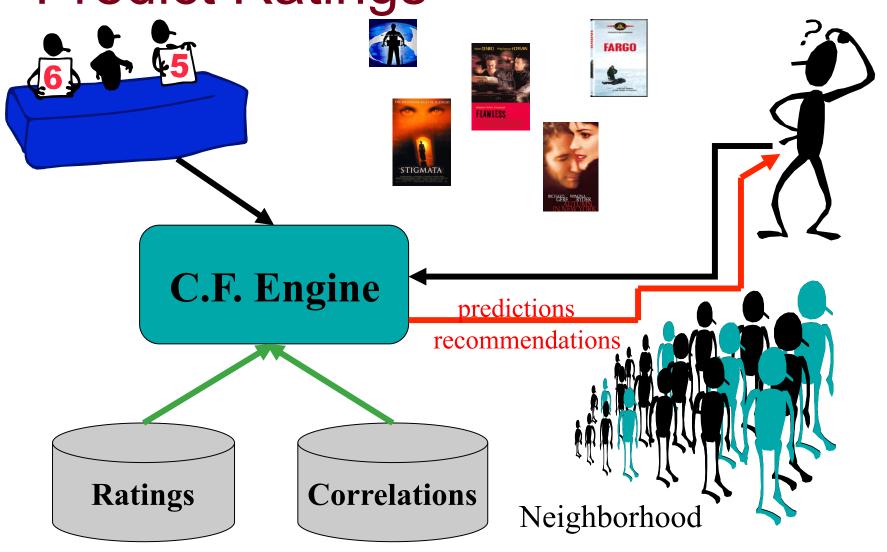
CF Classic: Request Recommendations



CF Classic: Identify Neighbors



CF Classic: Select Items; Predict Ratings



	Hoop Dreams	Star Wars	Pretty Woman	Titanic	Blimp	Rocky XV
Joe	D	A	В	D	?	?
John	A	F	D		F	
Susan	A	A	A	A	A	A
Pat	D	A		C		
Jean	A	\mathbf{C}	A	\mathbf{C}		A
Ben	F	A				F
Nathan	D		A		A	

	Hoop Dreams	Star Wars	Pretty Woman	Titanic	Blimp	Rocky XV
Joe	(D)	A	В	D	?	?
John	A	F	D		F	
Susan	A	A	A	A	A	A
Pat	D	A		C		
Jean	A	\mathbf{C}	A	C		A
Ben	F	A				F
Nathan	D		A		A	

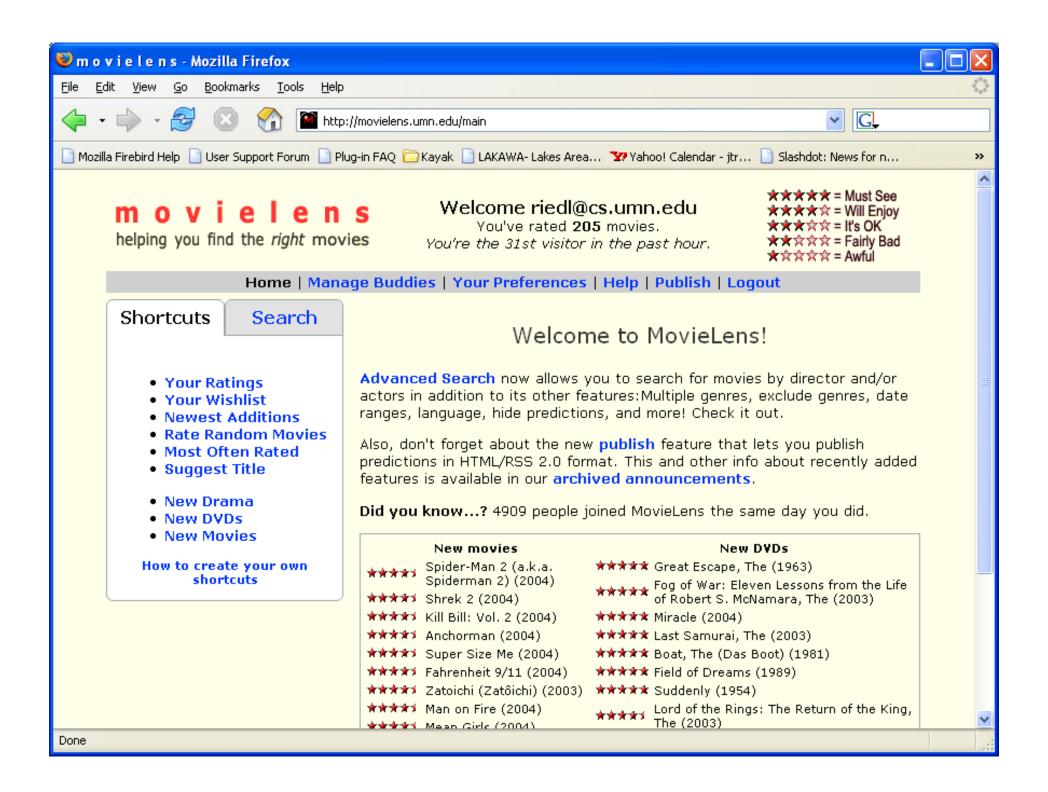
	Hoop Dreams	Star Wars	Pretty Woman	Titanic	Blimp	Rocky XV
Joe	(D)	A	(B)	D	?	?
John	A	F	D		F	
Susan	A	A	A	A	A	A
Pat	D	A		C		
Jean	A	C	A	\mathbf{C}		A
Ben	F	A				F
Nathan	D		A		A	

	Hoop Dreams	Star Wars	Pretty Woman	Titanic	Blimp	Rocky XV
Joe	0	A	B	D	?	?
John	A	F	D		F	
Susan	A	A	A	A	A	A
Pat	D	A		C		
Jean	A	\mathbf{C}	A	\mathbf{C}		A
Ben	F	A				F
Nathan	D		A		A	

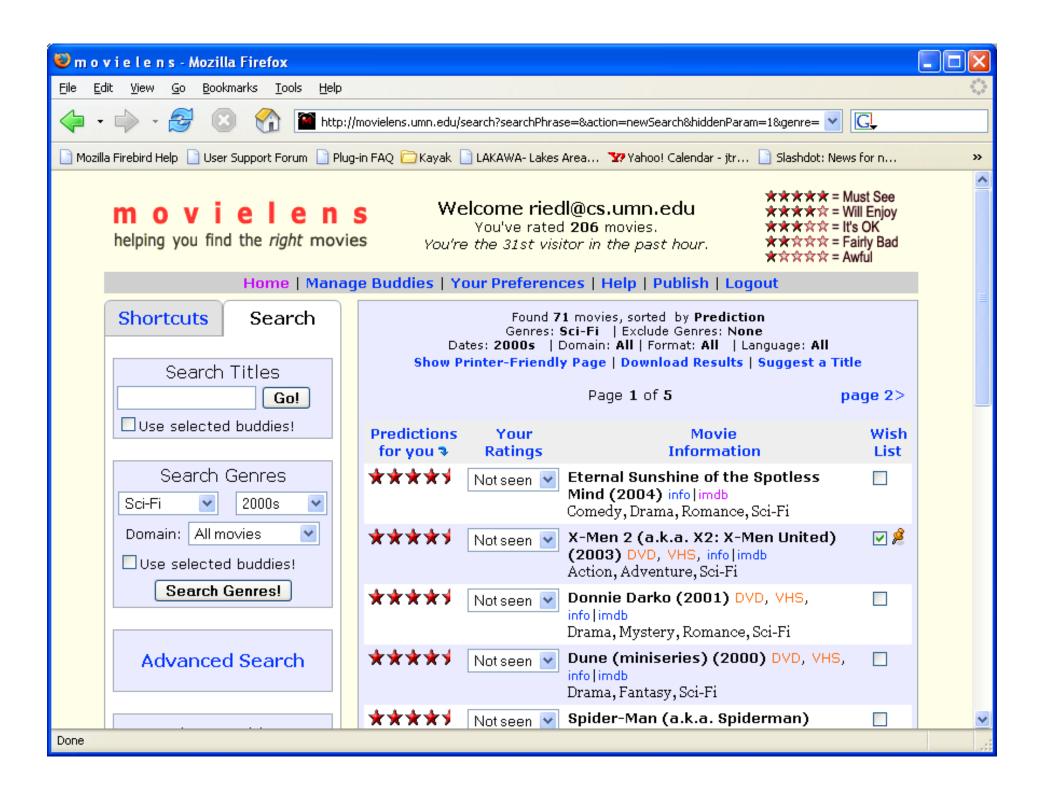
	Hoop Dreams	Star Wars	Pretty Woman	Titanic	Blimp	Rocky XV
Joe	Q	A	В	D	?	?
John	A	F	D		F	
Susan	A	A	A	A	A	A
Pat		A		C		
Jean	A	C	A	C		A
Ben	F	A				F
Nathan	D		A		A	

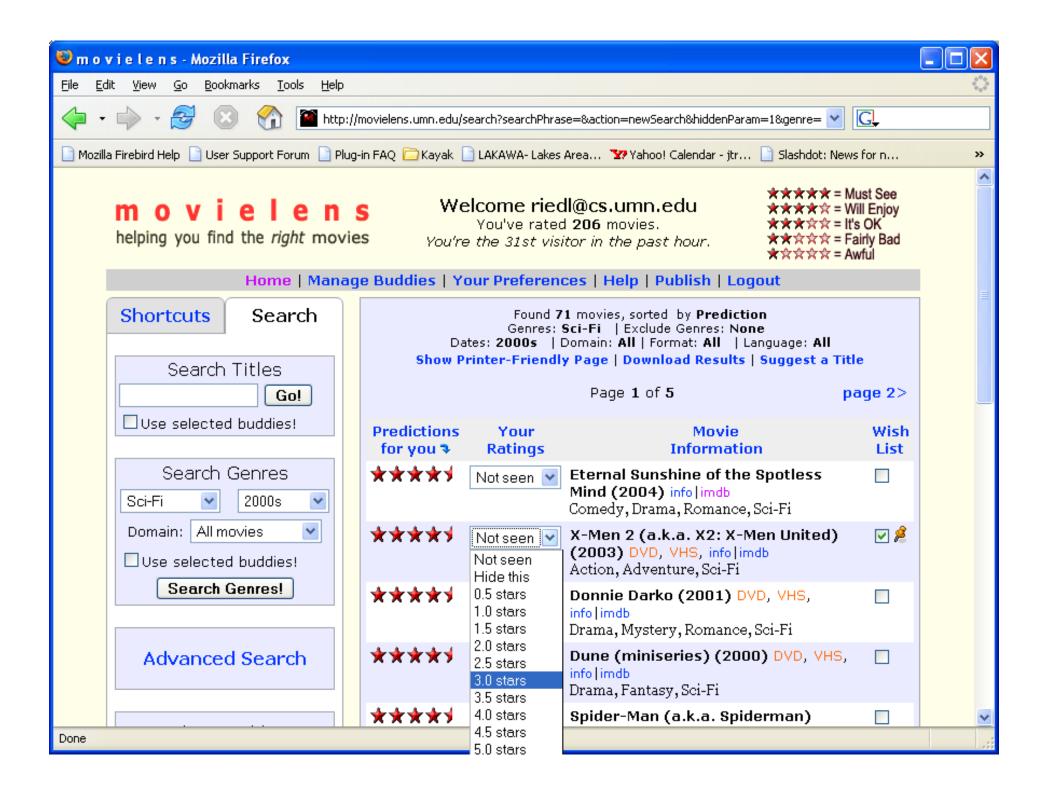
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Ben	F	A				F
Nathan	D		A		A	

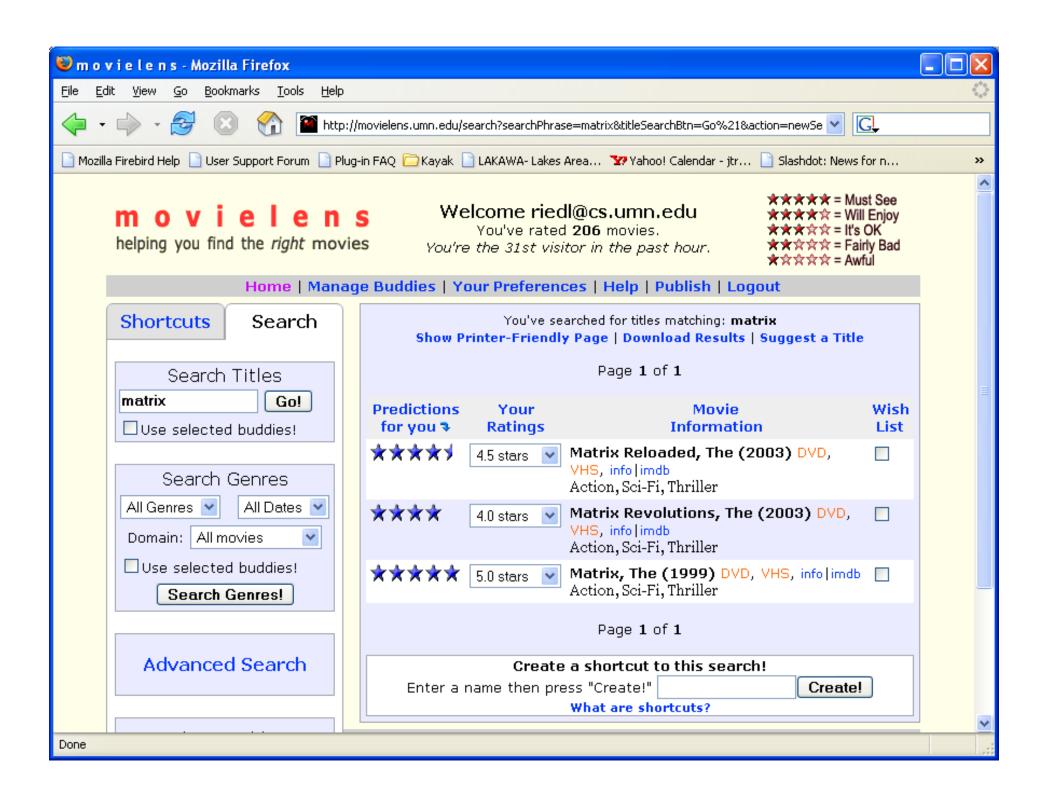
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Nathan	D		A		A	

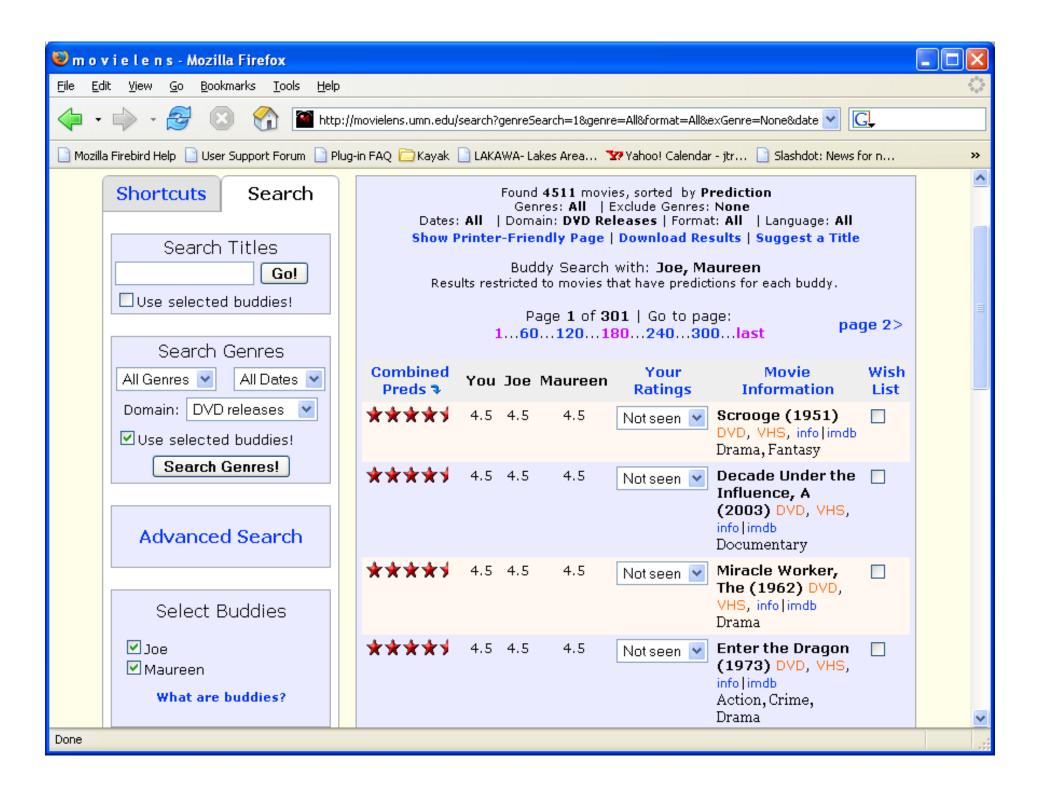




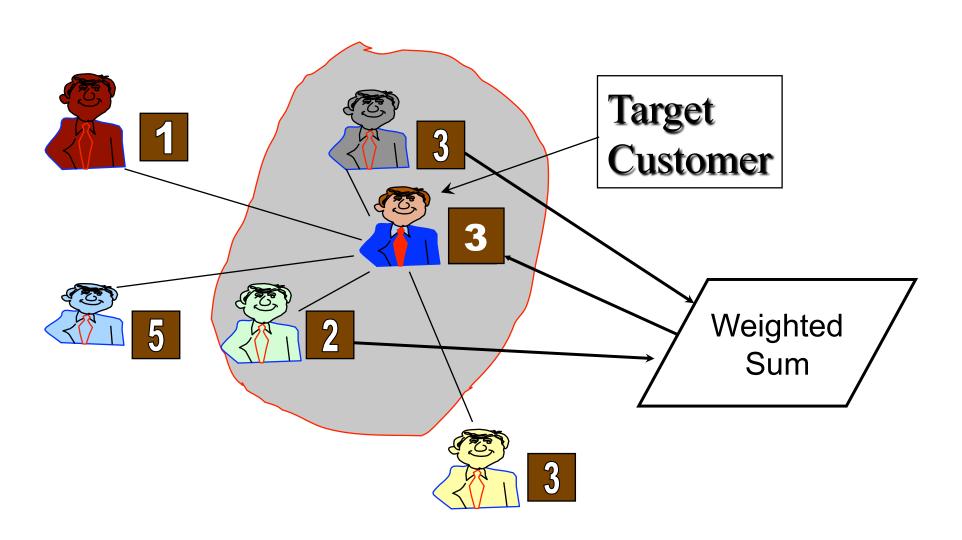








User-User Collaborative Filtering



Recommenders

- Tools to help identify worthwhile stuff
 - Filtering interfaces
 - E-mail filters, clipping services
 - Recommendation interfaces
 - Suggestion lists, "top-n," offers and promotions
 - Prediction interfaces
 - Evaluate candidates, predicted ratings

A Little Vocabulary

- Rating expression of preference
 - Explicit rating (direct from the user)
 - Implicit rating (inferred from user activity)
- Prediction estimate of preference
- Recommendation selected items for user
- Content attributes, text, etc.
- Collaborative using data from other users

Historical Challenges

Collecting Opinion and Experience Data

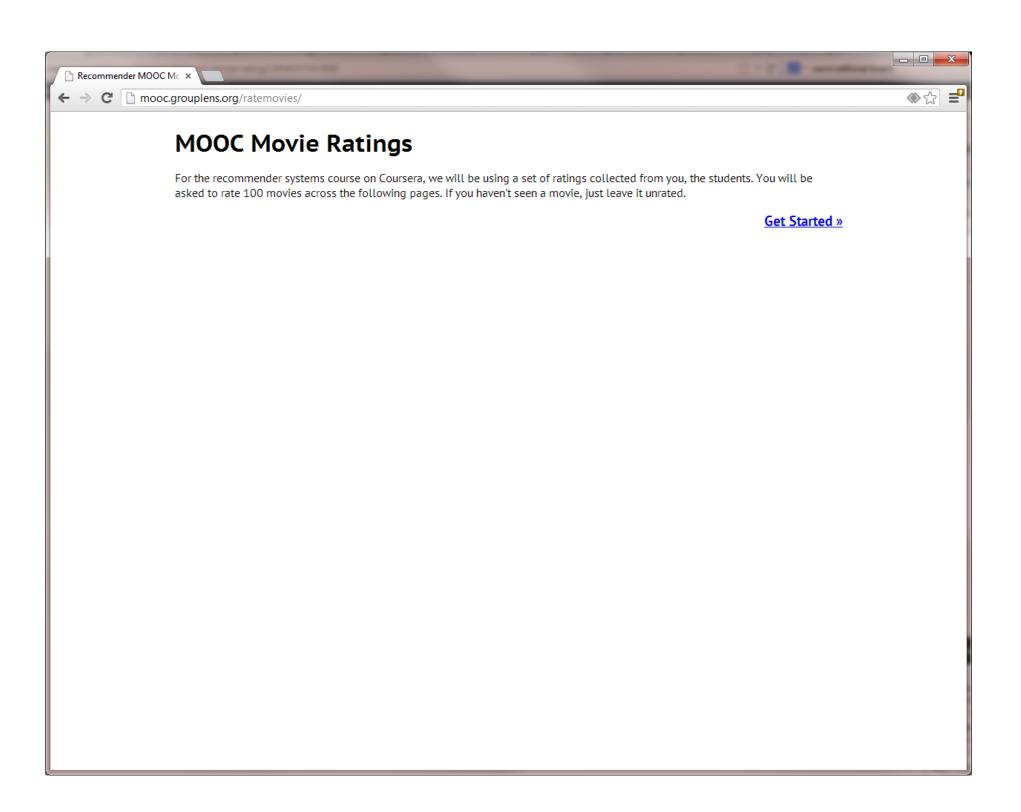
Finding the Relevant Data for a Purpose

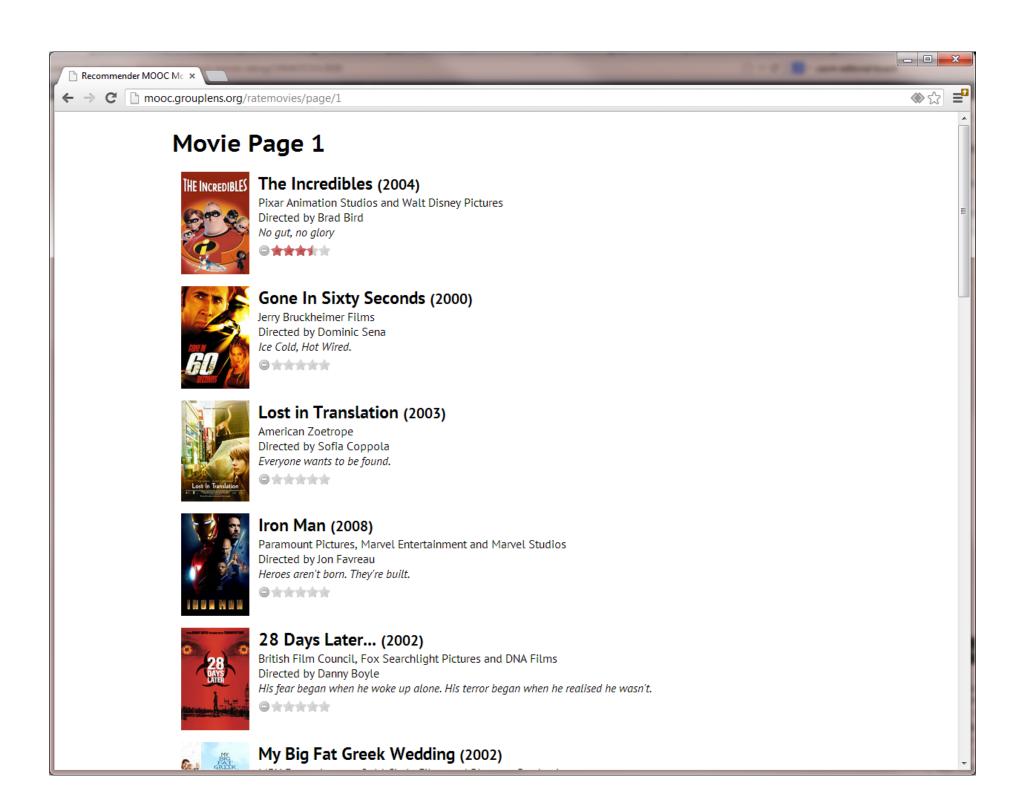
Computing the Recommendations

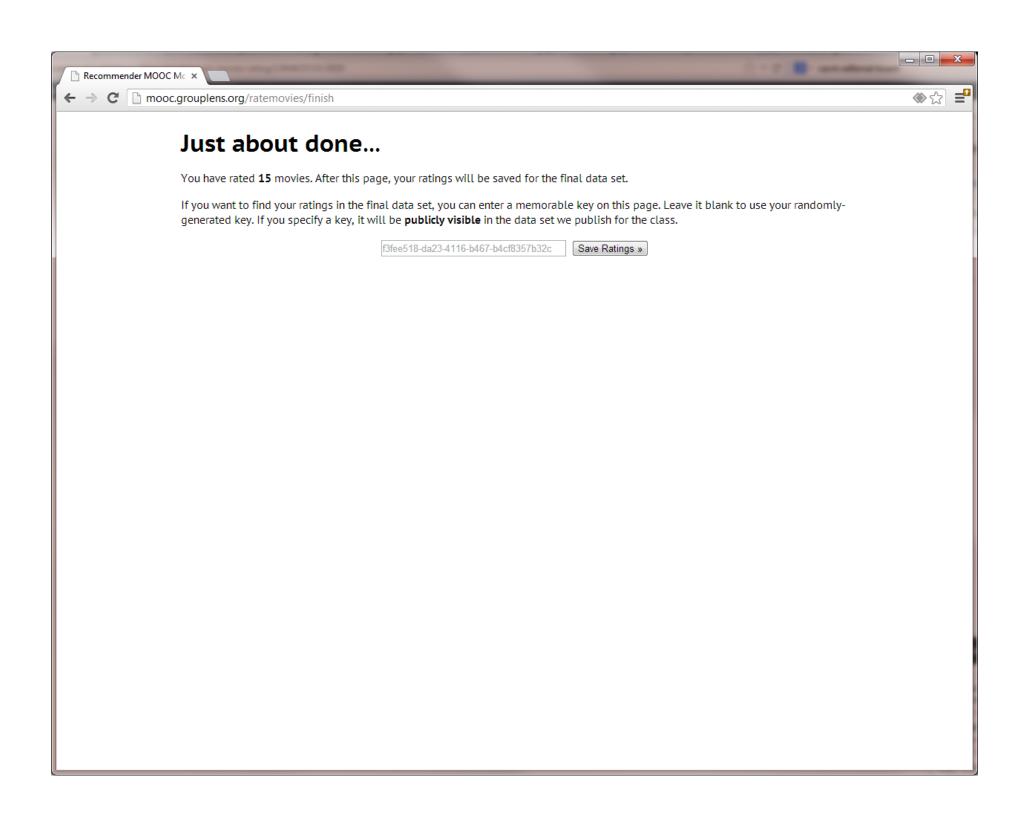
Presenting the Data in a Useful Way

Your First Assignment

- We are building a class ratings dataset using the MovieLens infrastructure
 - This will be used for several of the assignments
- Your assignment is to rate movies through our interface:
 - http://mooc.grouplens.org/ratemovies/







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 - Only rate movies you've seen
 - You can create a key to follow your preferences forward (or save the random key)

Moving Forward ...

- Next Lecture: Formal Course Introduction
- Rest of this Week
 - Technical details for programmers
 - Taxonomy
 - Tour of Recommenders in Amazon.com
- If not familiar with recommenders at all:
 - Spend some time using them (movielens.org)

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