3-1: Introduction to Content-Based Recommenders

Learning Objectives

- To understand the range and value of content-based approaches to recommendation
 - Pure information filtering systems
 - Case-based reasoning systems
 - Knowledge-based navigation systems
- To understand the strengths and drawbacks of content-based recommender systems

Basic Idea: Stable Preferences

- Let's consider some examples:
 - News I prefer stories on technology, University of Minnesota, Minnesota Vikings, restaurant reviews
 - Clothing I prefer cotton, blue, low-priced, casual
 - Movies I prefer Tom Hanks, Sandra Bullock, Woody Allen, Comedy
 - Hotels I prefer 24-hour front desk, room service, internet, pool

The key ideas

- Model items according to relevant attributes
- Model or reveal user preferences by attribute
- Voila! A Recommender

Content-Based Filtering

- Key concept: building a vector of attribute or keyword preferences
- Example: Krakatoa Chronicle

http://www.w3.org/Conferences/WWW4/Papers/93

Kamba, Bharat, and Albers (WWW '95)

Document URL:

http://homepark.cc.gatech.edu:8080/cgi-bin/oldHJnews.pl?Read=Read+Last+Edition+of+Krakatoa+Chronicl€

Tuesday, April 26, 1995

1/20

Krakatoa Chronicle

density

dense

tendency

general

sensitivty

not very sensitive

P: Peek an article, S: Save an article to a scrapbook

Technical static fuzzes Europe's telecom networks

(c) 1995 Bloomberg Business News

PARIS, April 26 (04-25-95) -With today's modems and telephone networks, sending data via telephone from a conference in Germany to your London office should be child's play, right?

Wrong.

Electric plugs differ from place to place. So do telephone connectors,

Interesting

(c) 1995 The News & Observer Publishing Co. dial tones and the latest transmission standards. As business traveler's can attest, a thicket of technical, regulatory and other hurdles must be cleared before Europe becomes a "plug-and-play" \$160 billion telecommunications market, as promised, by 1998.

It will have to be in place before doctors in Belgium and Germany

Plant Rogers executives resign as officers of Unitel Communications

(c) 1995 The News & Observer Publishing Co. (c) 1995 Bloomberg Business News

Toronto, April 26 (04-25-95) -Roger's Communications Inc., which spurned a majority stake in long-distance telephone company Unitel Communications Inc., said three of its executives resigned as officers and directors of Unitel.

The three executives who resigned are Ted Rogers, Phil Lind and David Gergacz.

No Comment

Rogers, who is president and chief executive of the cable-television operator, was Unitel's chairman, a post he had held since mid-January. Lind and Gergacz served as vice chairmen of Unitel.

Last week, Rogers Communications said it wouldn't exercise an option to buy a majority stake in Unitel. Toronto-based Rogers acquired the option in January from

$\frac{\mathbf{P}_{||}}{||}$ President will buy Wang stake and invest in a Shanghai plant

(c) 1995 Bloomberg Business News

TAIPEI, April 26 (04-25-95) — President Technology Inc., a Taiwan computer maker and

(c) 1995 The News & Observer Publishing Co. software services concern, said it purchased the 49% of a Taiwan software concern it didn't already own from Lowell, Mass.-based Wang Laboratories Inc. for NT\$365 million (\$14.4 million).

The company, 19% owned by Taiwan food processor President Enterprises Corp. (1216 TT), also said it plans to invest NT\$100 mil-

No Comment

lion in a computer monitor plant in Shanghai this year and double its capital to NT\$1.5 billion.

President Technology purchased

PR.R. Donnelley unit completes merger with Corporate Software

(c) 1995 The News & Observer Publishing Co. R.R. Donnelley & Sons said it has (c) 1995 Bloomberg Business News

Chicago, April 24 (04-24-95) -

Interesting

completed the merger of its

Creative Technology, Aztech expand share of sound card market

(c) 1995 Bloomberg Business News

Singapore, April 25 (04-24-95) -

Interesting

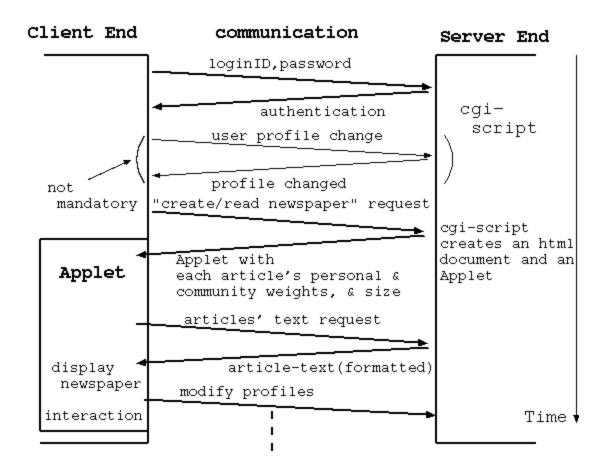
(c) 1995 The News & Observer Publishing Co. Creative Technology Ltd. and Aztech Systems Ltd. both expanded











Wide range of Possibilities

- User could build own profile (awkward)
 - But allowing user to edit a profile can be valuable
- Infer profile from user actions
 - Read, Buy, Click
- Infer profile from explicit user ratings
 - How to map from item preference to attribute preference
- We merge actions/explicit into infer from ratings (explicit and implicit)

How to build preferences?

- Let's start with the idea of a set of "keywords" that users may like, dislike, or not have an opinion on
- We could simply count the number of times the user chooses (or fails to choose) items with each keyword
- Or we can get more sophisticated
- More to come (future lecture) ...

How to use preferences

- Given a vector of keyword preferences
 - Do we just add up likes and dislikes?
 - Can we figure out which keywords are more and less relevant?
- Forward reference: TFIDF

Content-Based Recommenders

- Our assignments will be based on this model
 - Hand exercises: building a profile and using it to predict a few cases
 - Programming exercise: building a content-based recommender

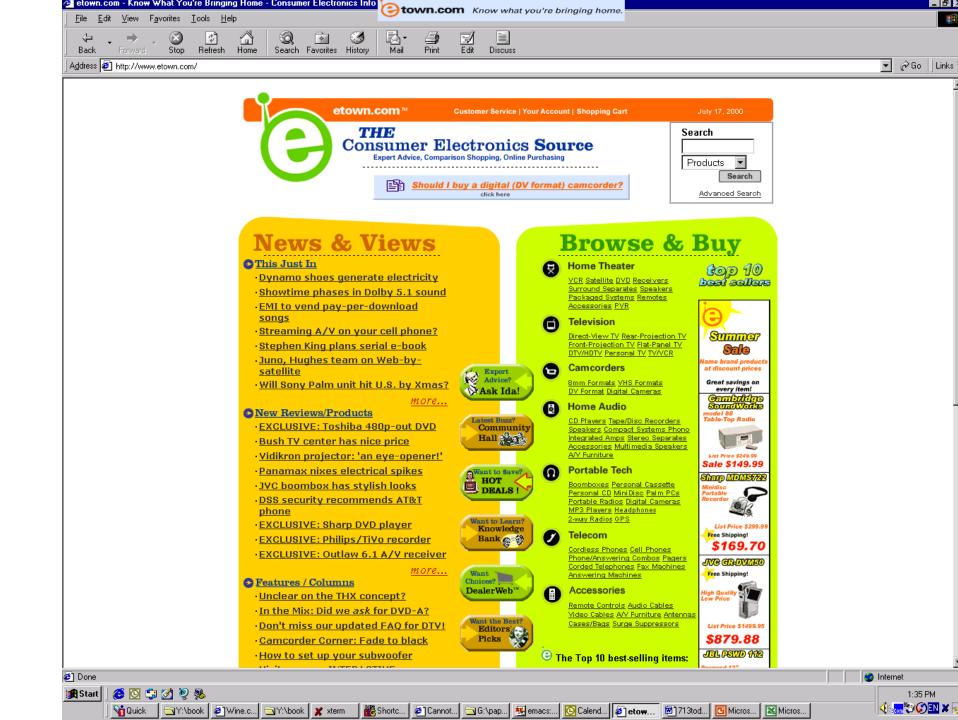
But first, a few other approaches ...

Case-Based Recommendation

- The concept:
- Structure a database of cases around a set of relevant attributes (e.g., camera price, zoom, pixels)
- Query based on an example or attribute query, and retrieve relevant cases
- Open issue: Many ways to structure interaction

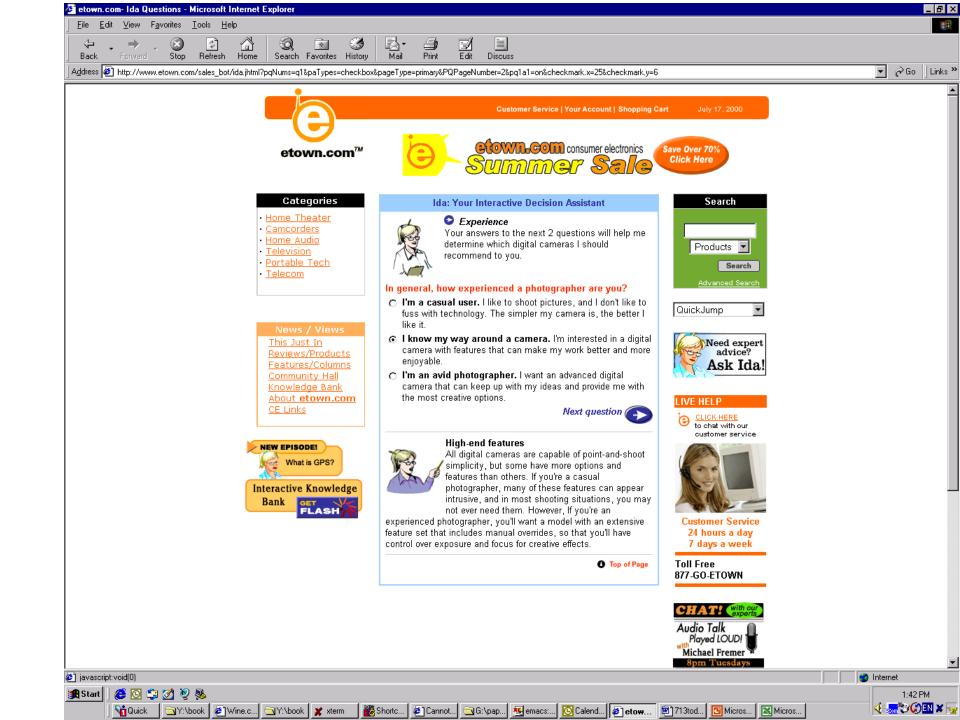
etown's Ask Ida

- No longer exists (old screenshots)
- Uses an interview process to elicit preferences over attributes
- Uses preferences to recommend products
- Uses recommendation as a point to elicit further preferences
- Note: not intended as permanent preferences – just transactional

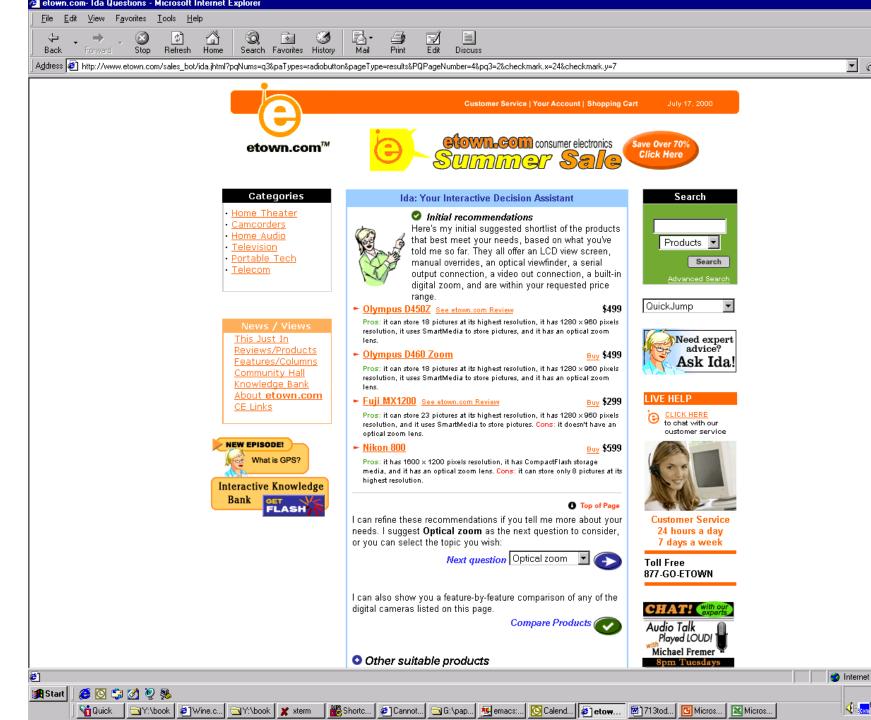






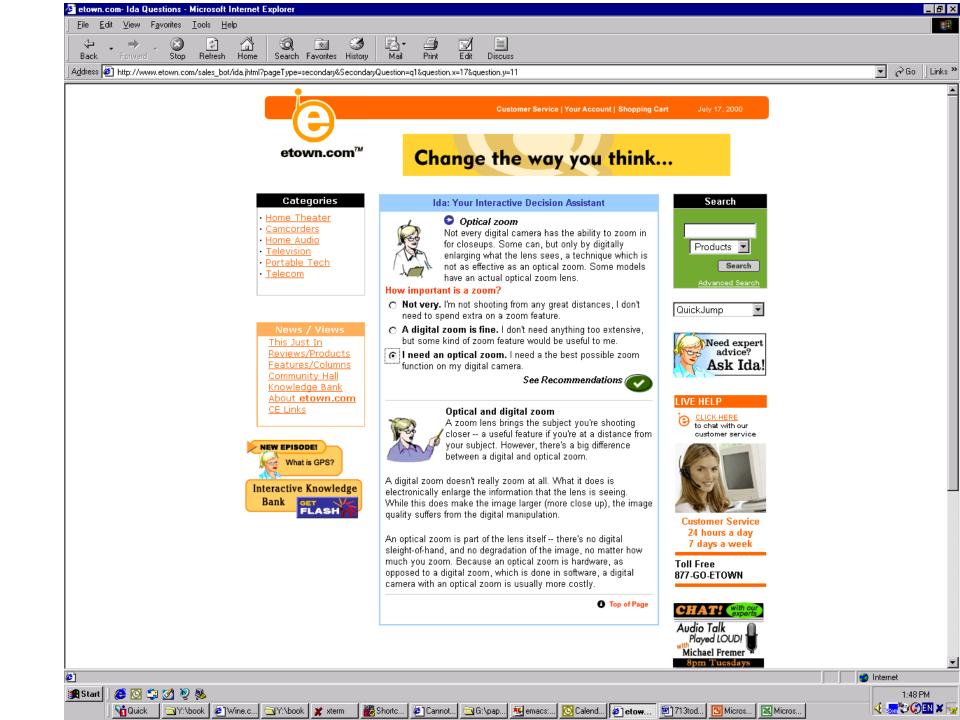




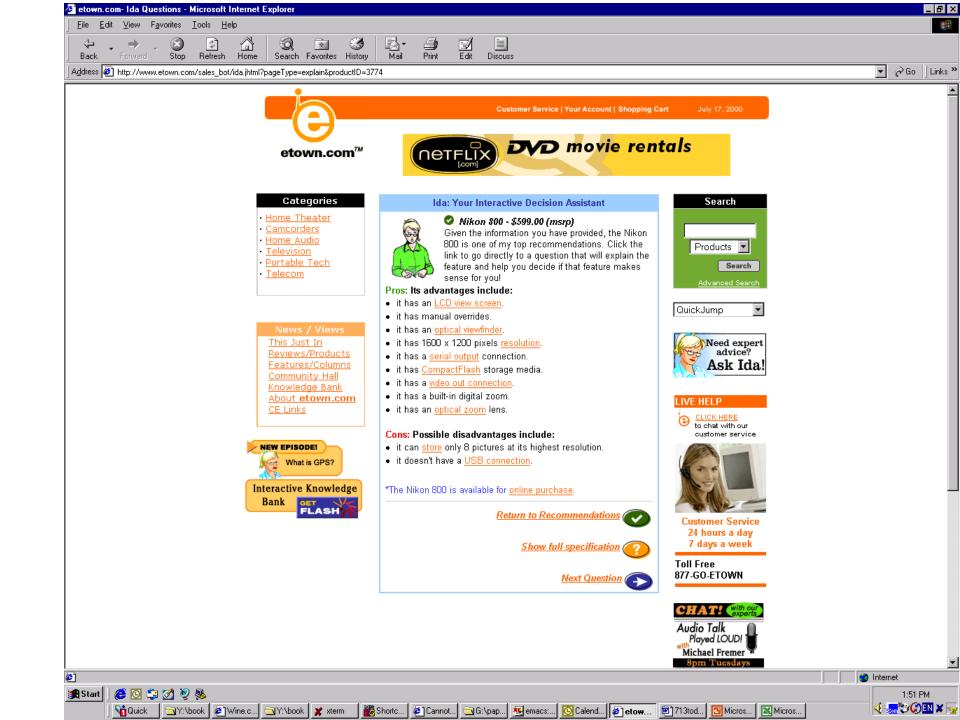


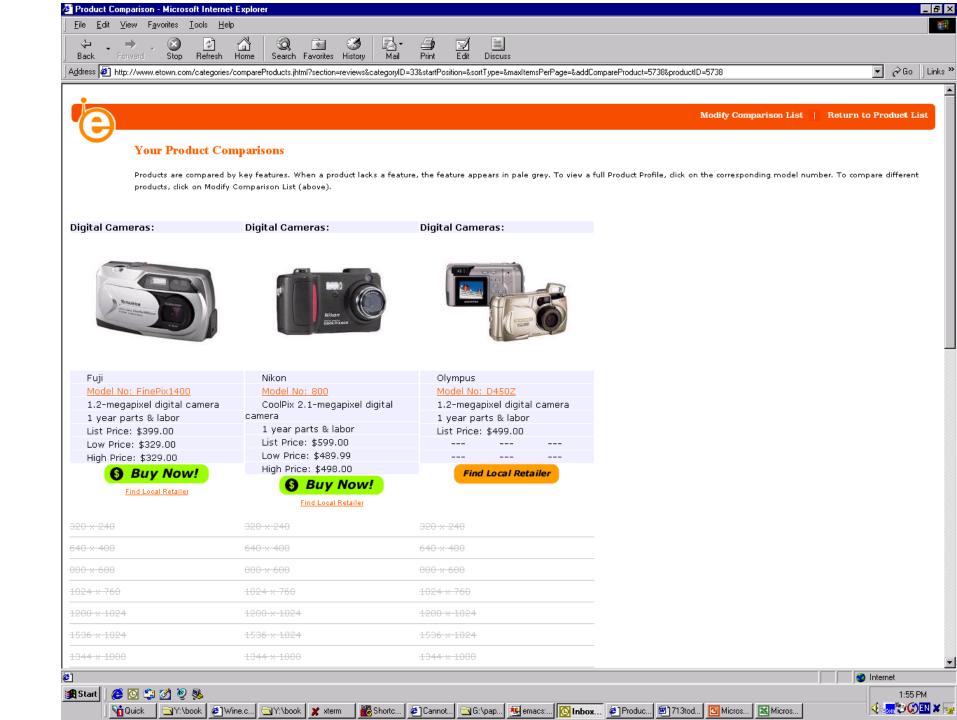
∂Go

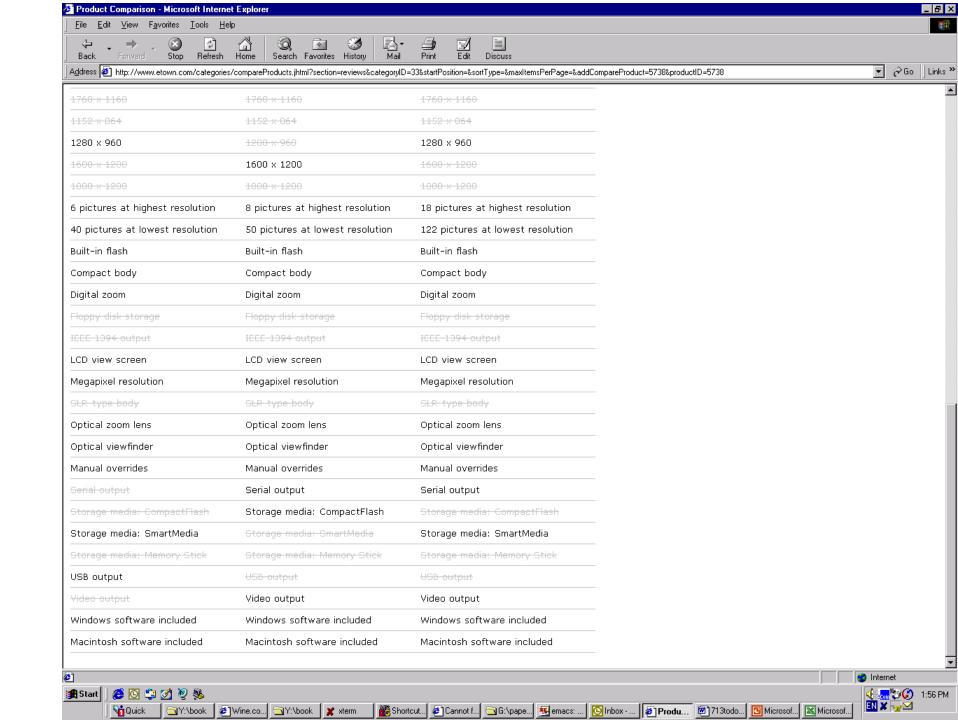
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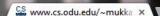






Knowledge-Based Recommender

- Case-Based Example with Navigation Interface
- FindMe Systems (e.g., Entrée)









www.cs.odu.edu/~mukka/cs795sum10dm/Lecturenotes/Day6/burke-elis00.pdf





The Los Angeles restaurant you chose is:

Chinois On Main

2709 Main St. (bet. Rose Ave. & Ocean Park Blvd.), Santa Monica, 310-392-9025

Pacific New Wave

\$30-\$50

Extraordinary Decor, Extraordinary Service, Near-perfect Food, Hip Place To Be, On the Beach, Great for People Watching, Parties and Occasions, Weekend Brunch, Weekend Lunch, Fabulous Wine Lists

We recommend:

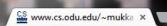
Yoshi's Cafe

3257 N. Halsted St. (Belmont Ave.), Chicago, 312-248-6160

Asian, Japanese, French (New)

\$30-\$50

Extraordinary Decor, Extraordinary Service, Near-perfect Food, Need To Dress, Prix Fixe Menus, Quiet for Conversation, Very Busy - Reservations a Must, Romantic, Good Out of Town Business, Fabulous Wine Lists, Game, Parking/Valet









www.cs.odu.edu/~mukka/cs795sum10dm/Lecturenotes/Day6/burke-elis00.pdf





Figure 3: Navigation using the "Less \$\$" tweak

More Generally

- Case-Based Approaches (Knowledge, Database, etc.) are often most helpful for ephemerally-personalized experiences
 - Shopping suggest similar relevant items
 - Compare with collaborative suggest items that are co-purchased or co-browsed
 - Content suggest similar stories
- Case-Based recommendations are often easier to explain to the user

Challenges and Drawbacks

- Content-Based Techniques in general ...
 - Depend on well-structured attributes that align with preferences (consider paintings)
 - Depend on having a reasonable distribution of attributes across items (and vice versa)
 - Unlikely to find surprising connections (e.g., chili peppers or lemon with chocolate)
 - Harder to find complements than substitutes

Some take-away lessons

- Many ways to recommend based on content (product attributes)
 - Long-term: build profile of content preferences
 - Shorter-term: build database of cases; navigate
- Content-based techniques work without a large set of users (but need item data)
- Good at finding substitutes; good at helping navigate for a purchase; good explainability

Moving Forward

- Next Lectures
 - For programmers: Introduction to LensKit
 - For everyone: deeper dive into content profiles, content retrieval and filtering
- Later this Module
 - Programming deep dive; guest lectures on casebased and knowledge-based; survey of tools for content recommending

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