# 1-4(b): Taxonomy of Recommender Systems (part 2 of 2)

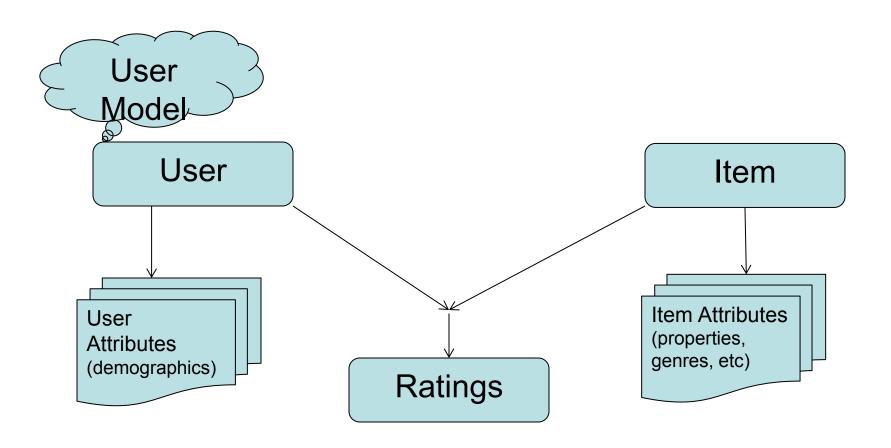
### Recommendation Algorithms

- Non-Personalized Summary Statistics
- Content-Based Filtering
  - Information Filtering
  - Knowledge-Based
- Collaborative Filtering
  - User-User
  - Item-Item
  - Dimensionality Reduction
- Others
  - Critique / Interview Based Recommendations
  - Hybrid Techniques

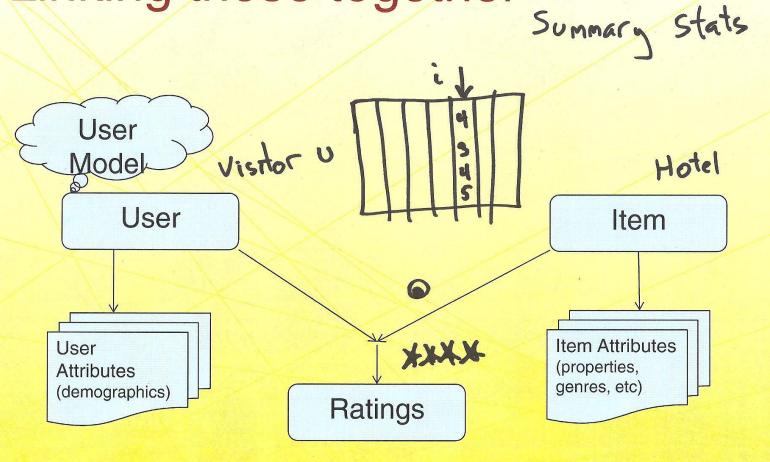
# From the Abstract to the Specific

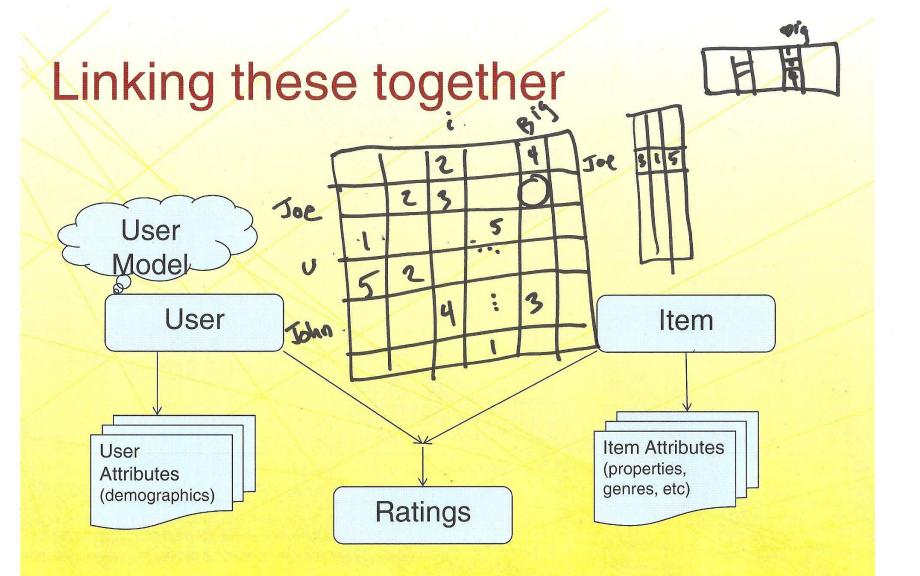
- Basic Model
  - Users
  - Items
  - Ratings
  - (Community)

# Linking these together



Linking these together





Linking these together Summary Stats User Vistor U Model Hotel User Item XXXX Item Attributes User (properties, **Attributes** genres, etc) (demographics) Ratings

# Non-Personalized Summary Stats

- External Community Data
  - Best-seller; Most popular; Trending Hot
- Summary of Community Ratings
  - Best-liked

- Examples
  - Zagat restaurant ratings
  - Billboard music rankings
  - TripAdvisor hotel ratings

# Content-Based Filtering

- User Ratings x Item Attributes => Model
- Model applied to new items via attributes
- Alternative: knowledge-based
  - Item attributes form model of item space
    - Users navigate/browse that space
- Examples
  - Personalized news feeds
  - Artist or Genre music feeds

#### Personalized Collaborative Filtering

- Use opinions of others to predict/recommend
- User model set of ratings
- Item model set of ratings
- Common core: sparse matrix of ratings
  - Fill in missing values (predict)
  - Select promising cells (recommend)
- Several different techniques

# Collaborative Filtering Techniques

- User-user
  - Select neighborhood of similar-taste people
    - Variant: select people you know/trust
  - Use their opinions
- Item-item
  - Pre-compute similarity among items via ratings
  - Use own ratings to triangulate for recommendations
- Dimensionality reduction
  - Intuition: taste yields a lower-dimensionality matrix
  - Compress and use a taste representation

#### Note on Evaluation

- To properly understand relative merits of each approach, we will spend significant time on evaluation
  - Accuracy of predictions
  - Usefulness of recommendations
    - Correctness
    - Non-obviousness
    - Diversity
  - Computational performance

# Other Approaches

- Interactive recommenders
  - Critique-based, dialog-based
- Hybrids of various techniques

# Moving Forward

- Next Lecture: A Tour of Amazon.Com, organized by our taxonomy
- Then, you should be able to:
  - Analyze a recommender application on your own
- Course Structure:
  - We step through the recommendation algorithms, with six major modules
  - Related topics intermingled

# 1-4: Taxonomy of Recommender Systems