



# Integrating Quality Criteria in a Fuzzy Linguistic Recommender System for Digital Libraries

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- Introduction.
- Proposed system.
- Conclusions.

# Introduction



- Web: Main source of information generation and transmission.
- We focus on an academic environment: University Digital Libraries (UDL).



## Information Access Problems

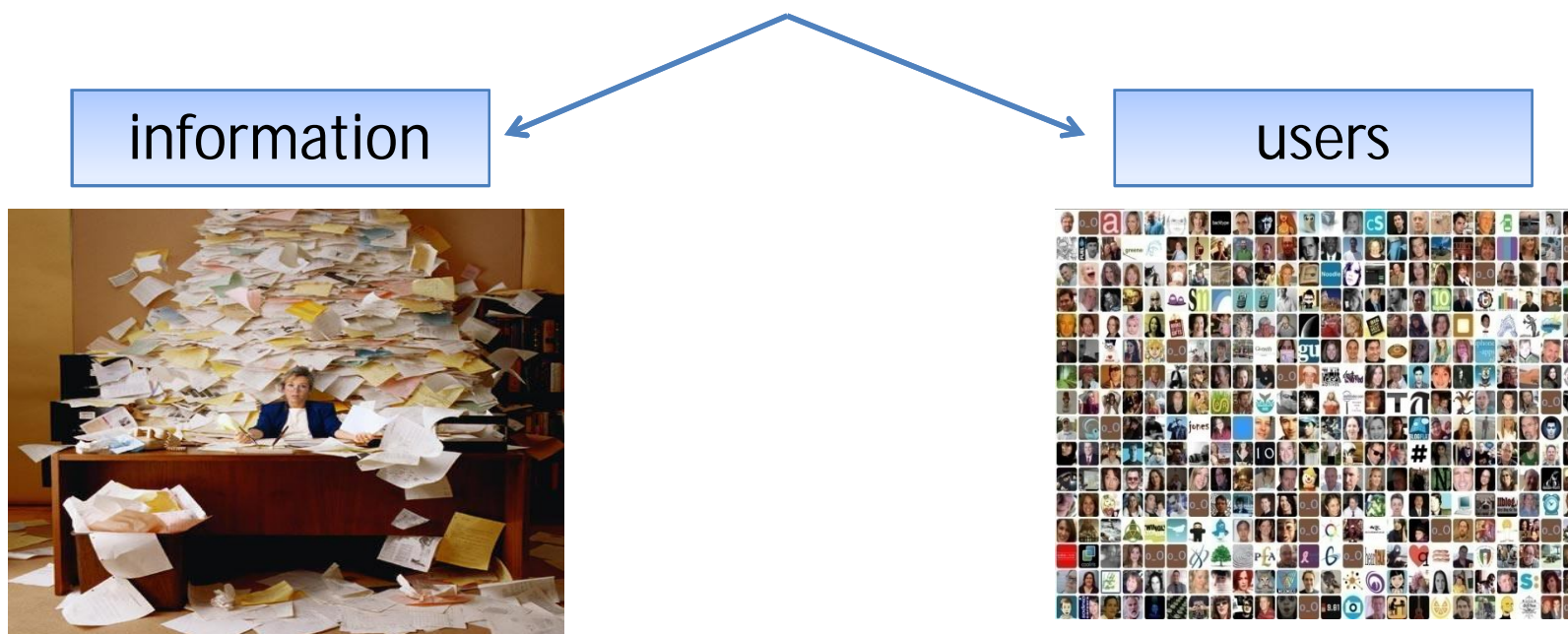
- Need for automatic search systems and access to the information in the Web:
  - Recommender Systems (RecSys): They aid users in the information access process through prediction and item recommendation that can be interesting for them → users' profile.



# Introduction



- **Main problem** in the Web: exponential and uncontrolled:



- **Consequence:** the users of UDL still having serious difficulties to access to relevant information.

# Introduction



## Proposed solution

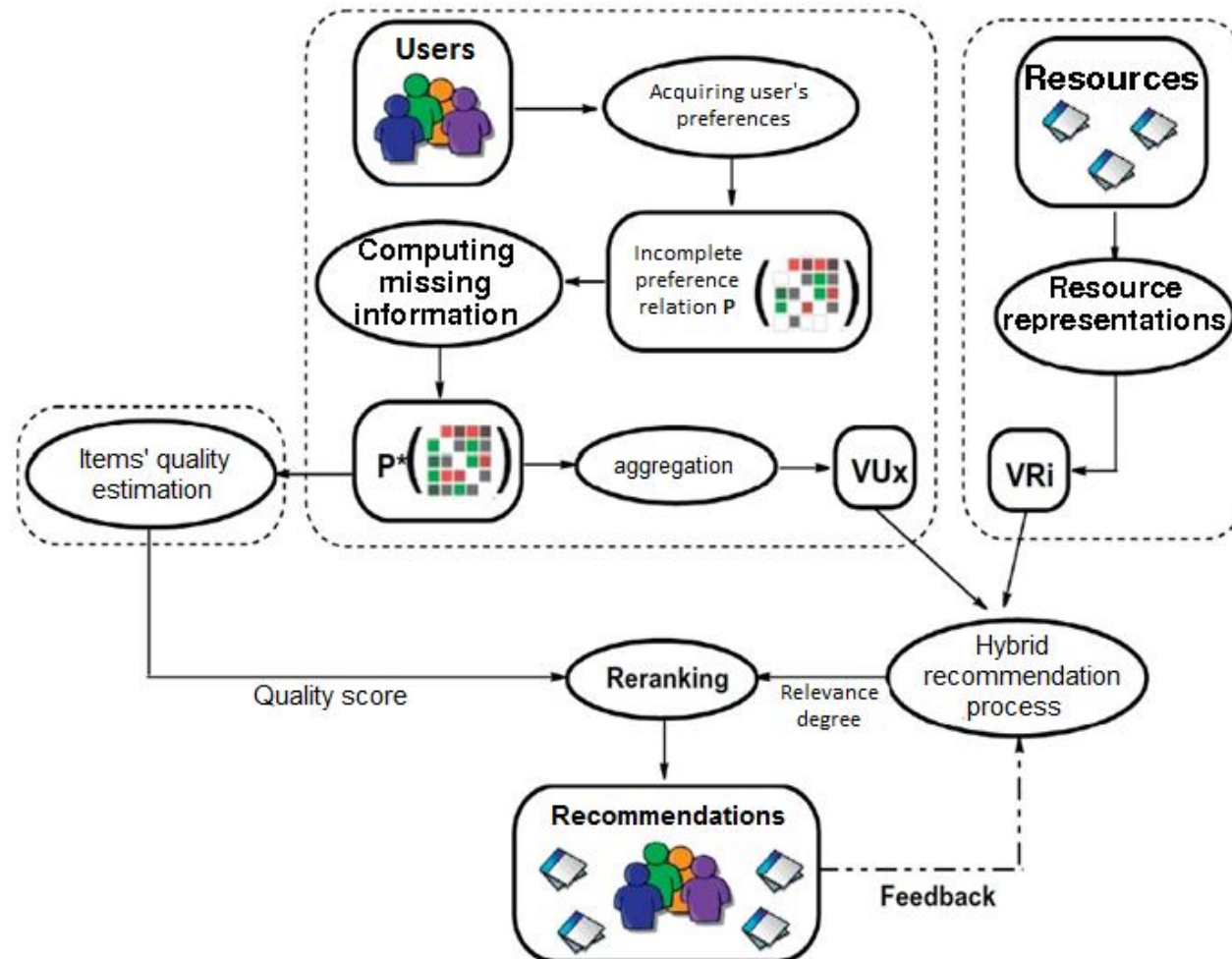
- We split the process of generating recommendations in **two phases**:
  1. Identify **relevant** resources.
  2. Identify valid resources from a **quality** point of view.
- **Hybrid recommendation** → *Switched hybrid RecSys*: To alternate between a content-based scheme and a collaborative one depending on the number of existing ratings.
- To add the **Re-ranking module** which combines the estimated relevance degree with the quality of the item.
- To adopt a **multi-granular fuzzy linguistic modeling**.



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# Description of the proposed system



# Proposed system : representation of Information



- We use different sets of labels selected from a **linguistic hierarchy**.
- Concepts assessed:
  1. **Relevance degree** of a discipline with respect to a resource scope, which is assessed in  $S_1$ .
  2. **Similarity degree** among resources or among users, which is assessed in  $S_2$ .
  3. **Predicted relevance degree** of a resource for a user, which is assessed in  $S_3$ .
  4. **Satisfaction degree** expressed by a user to evaluate a recommended resource, which is assessed in  $S_4$ .
  5. **Preference degree** of a resource regarding another one, which is assessed in  $S_5$ .
- We use 5 labels to  $S_1$  y  $S_5$ , and 9 to  $S_2$ ,  $S_3$  y  $S_4$ .



# Proposed system : representation of Information



## Resources representation

- To represent the resource scope, we use a **vector model**.
- We use a classification of by 25 disciplines.
- A resource  $i$ , is represented as:

$$VR_i = (VR_{i1}, VR_{i2}, \dots, VR_{i25})$$

- where  $VR_{ij}$  ( $S_1$  labels) shows the importance degree of discipline  $j$  regarding to resource scope  $i$ .

# Proposed system : representation of Information



## User profiles

1. To acquire **users' preferences** over the 5 most representative resources.
  - It is enough for users to provide a row of the relation and the system will complete the relation ( $S_5$  Labels).
2. To calculate user **resource preference degrees** over each considered resource  $\rightarrow$  arithmetic mean.
  - Now we can obtain the **user preference vector** as the aggregation of vectors representing selected resources characteristics, weighted through preference degrees.

# Proposed system : Recommendation scheme



## Hybrid scheme

- It allows us to face the cold start problem.
- Similarity measures: standard *cosine measure*, but defined in a linguistic context ( $S_2$  labels).
- *Content-based* approach: when a new resource is inserted.
- Collaborative approach: when a new user is inserted.
- Then, the relevance of a resource for a user is estimated ( $S_3$  labels).

## Proposed system : Quality estimation



- Idea: If a resource is usually preferred over others that show a certain quality.



Probability of this resource be preferred over other having been selected

- At the stage of completing the incomplete preference relations we count the number of times a resource  $i$  is chosen to be shown among the outstanding resources,  $(s_i)$  is the total of times the resource  $i$  has been selected and the total number of times  $i$  has been preferred over other  $(p_i)$ :

$$q(i) = p_i/s_i$$

- **Advantages:** It avoids to collect additional information about users and to increase the complexity.

## Proposed system : Reranking



- We aggregate the estimated relevance with the quality score obtained.
- We use a **multiplicative aggregation** and we normalize it in the range of the label set  $S_3$ .
- **Advantages:** ease of application and good results obtained.

## Proposed system: Feedback



- The activity of generating recommendations is completed with this phase.
- Users provide the system with their satisfaction ratings about the items received ( $S_4$  labels).



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# Conclusions



- We have addressed the recommendations process from two perspectives:
  1. Find relevant resources.
  2. Resources of good quality.
- We have presented a **hybrid fuzzy linguistic recommender system applied to a UDL**.
- We performed online studies → satisfactory results.
- **Future works:**
  - Techniques for automatic resource representation.
  - Incorporate new techniques in the recommendation process.





Any question?