



Tutorial on Medical Image Retrieval - user interaction -

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Overview

- **Usability** issues
 - Interfaces, query speed, etc.
- **Relevance feedback**
 - AND, OR, NOT, XOR
 - Text retrieval
 - Positive and negative
 - Problems
 - Several steps
- **Long-term** analysis of user behavior
 - Log file analysis
 - Market basket analysis



Usability issues

- Often neglected in research projects
- Several dimensions:
 - Interactivity
 - Interfaces
 - Explain the output to the user
 - Different interfaces for novice, expert users
 - Standards for system use
- **Nielsen: Usability Engineering**



Usability: Query speed (Nielsen)

- 10 s: maximum to keep the attention focused on the dialogue
 - 1 s: maximum for **interactive** working
 - 0.1 s: **instantaneous** response
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- We need to aim for 1 s
 - Often systems state >3 minutes
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- Features etc. can be **precalculated**
 - Much literature on optimizing the query speed



Query starting point

- **Example image(s)**
 - Internal images (precalculated features)
 - External images
- Sketch
- Marking of **regions** in images
- Selection of pre-segmented regions
- **Text**
 - Free text
 - Database fields



User interfaces

- **Standards** are evolving, similar to web-based navigation (google for images)
- Many of the shown interfaces were similar
 - N=20-50 images on screen
 - Marking of images/regions for feedback
 - Function for random images
- **Innovative** interfaces:
 - 3D interfaces
 - Interaction with a data glove
 - Basically for sketching
 - Manual grouping of images to define a distance measure

Example interface: medGIFT

Query image

Link to casimage

Image	Diagnosis	Similarity	Dropdown
	Bilateral basal pneumonia Query Image (1.000000)	1.000000	Neutral
	Pneumonie basale bilatérale	1.000000	Neutral
	Pneumonie basale bilatérale	1.000000	Neutral
	Pneumopathie interstitielle l...	0.497142	Neutral
	Lymphocytic interstitial pneu...	0.497142	Neutral
	Extrinsic allergic alveolitis...	0.493630	Neutral
	Alvéolite allergique extrinsè...	0.493630	Neutral
	Alvéolite allergique extrinsè...	0.493630	Neutral
	Sarcoïdose stade I et II	0.492040	Neutral
	Sarcoïdose stages I and II	0.492040	Neutral
	Sarcoïdose stade I et II	0.492040	Neutral
	Sarcoïdose stade I et II	0.492040	Neutral
	Pneumonia in the left lower l...	0.489103	Neutral
	Pneumonia du lobe inférieur g...	0.489103	Neutral
	Pneumonia in the left lower l...	0.489103	Neutral
	Pneumonia du lobe inférieur g...	0.489103	Neutral
	Pneumonia in the left lower l...	0.488424	Neutral
	Pneumonia du lobe inférieur g...	0.488424	Neutral

Diagnosis

Similarity score



Relevance feedback

- **Query refinement** of a query with new and/or more example images from the result set
- Sometimes clearer formulation possible, of what the user wants
 - AND, OR, NOT, XOR
- Most often: images marked as **relevant or non-relevant**
 - Sometimes more gradual
- Two ways of calculating this
 - Separate queries for every image (\sim or)
 - Creation of a «pseudo-image» and one query (\sim and)



Positive and negative feedback

- Studies on strategies for relevance feedback
 - Positive feedback often a reordering of top results or one new query with a single image
 - Images already have much in common
 - **Negative** feedback is key to good results
 - Really new images are retrieved
 - Much more information is supplied
 - Problem with **two much negative** feedback!
 - Images with small number of features are returned
 - As much feedback as possible delivers best results

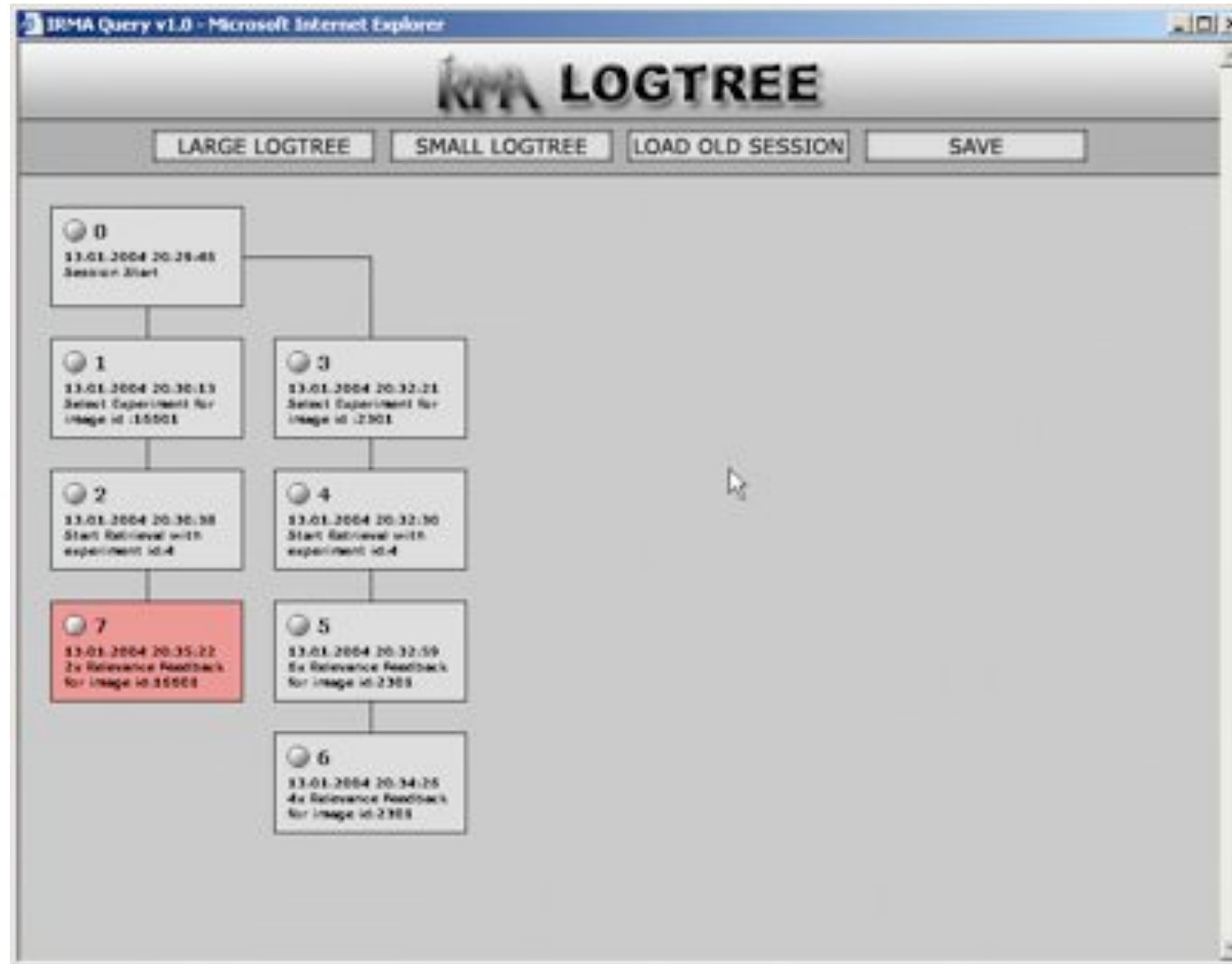


Rocchio Feedback (1960s)

- Problem with too much negative feedback also in text retrieval
- Solution: **Separately weighting** positive and negative parts of feedback
 - Often positive=0.65, negative=0.35

$$tf_j = \frac{\alpha}{n_1} \sum_{i=1}^{n_1} Rel_i \cdot tf_{ij} - \frac{\beta}{n_2} \sum_{i=1}^{n_2} Rel_i \cdot tf_{ij}$$

Storage of an interaction tree

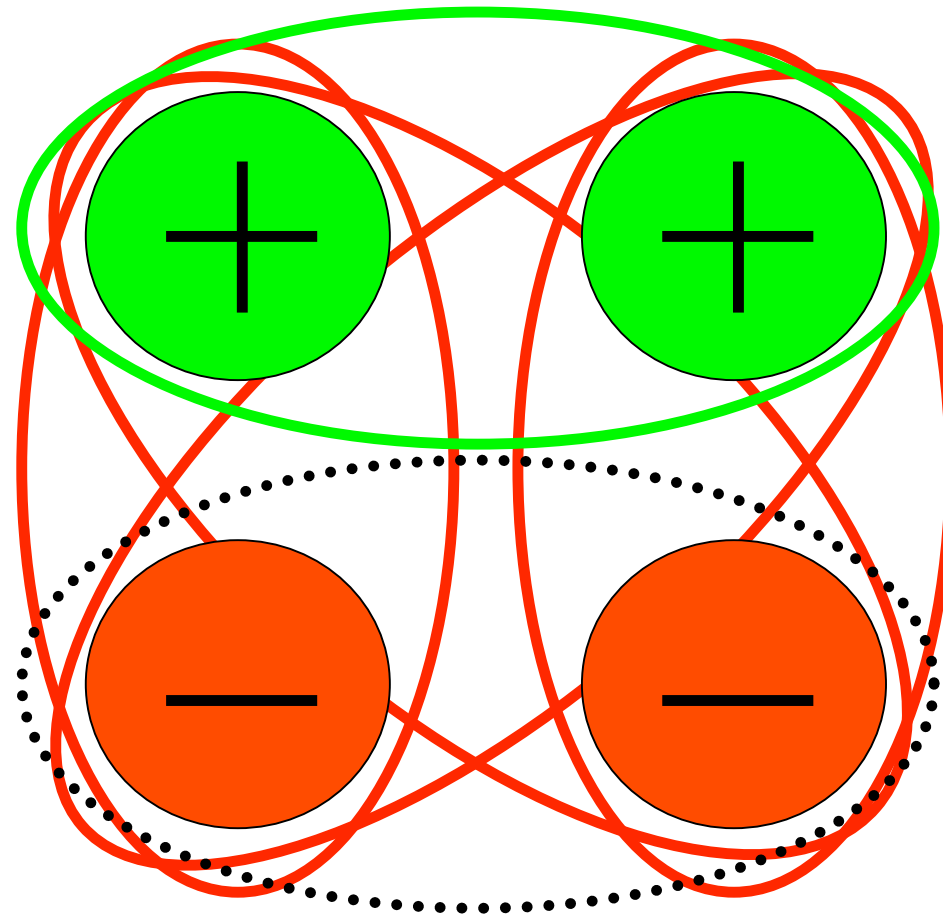




Market basket analysis

- Artificial intelligence problem
- Articles bought together by consumers in a **supermarket**
 - Extremely large data sets exist
 - Efficient calculation is the main goal
- How does a change in layout change what the consumers buy, what is bought together anyways
- Some similarities with image retrieval
 - Which features perform well based on what the users do

Combination of images from usage log files



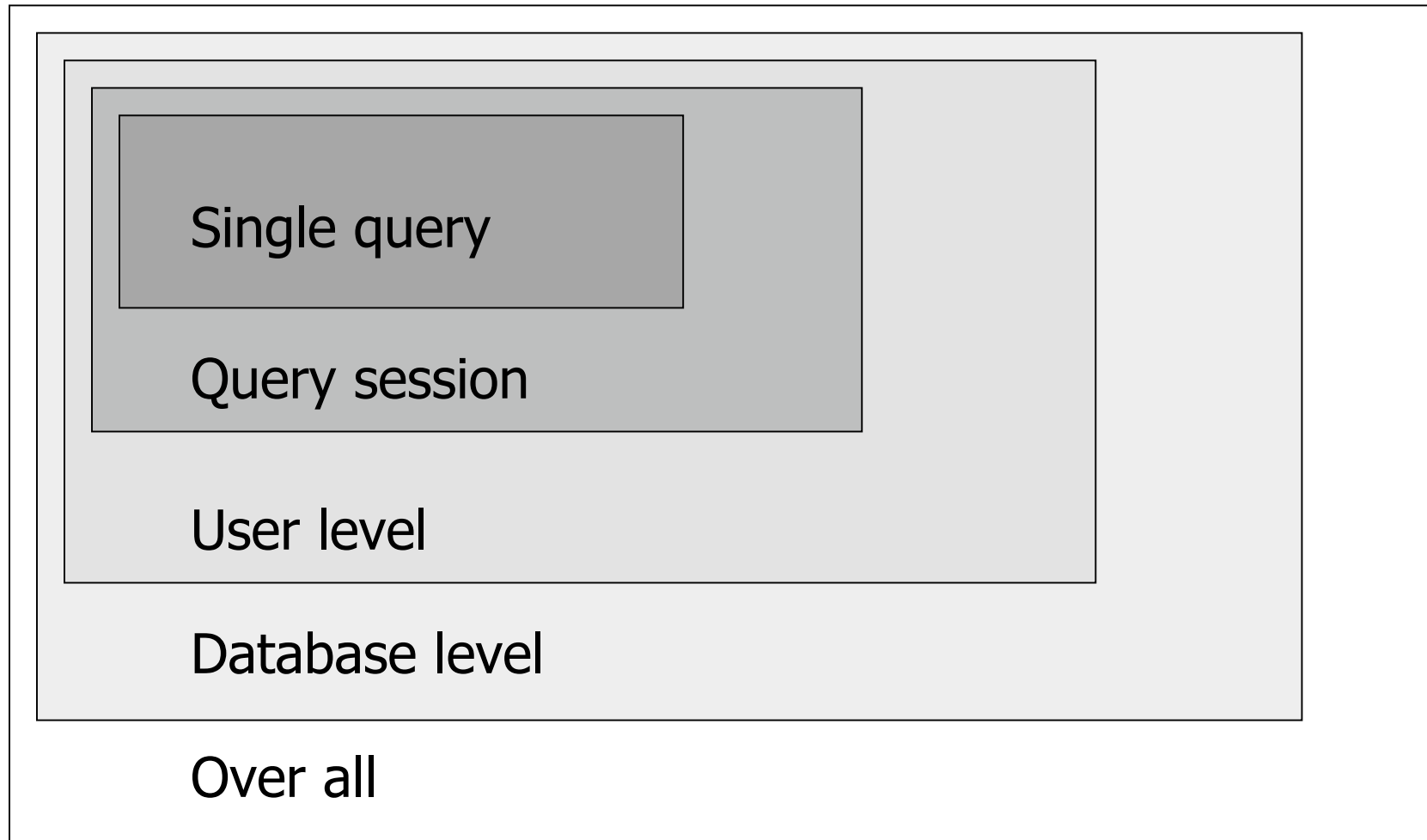


Learning from log files

- Large log files of user interaction exist from web demos
- Find out the **images marked together**
 - To return them for a similar query
- Better: Find out which **features** these images have in common
 - Calculate the important for each feature
 - Include it into the distance measure
 - Include it into the feature weighting
 - Can be an additional factor
- Goal: system gets **better over time** the more it is used
- Learning over several databases does not work too well



Learning over several scales





Conclusion

- Interaction is the **key** for information retrieval success
 - Interfaces, speed, ...
- System needs to be able to get as much information as possible on user goals
- When logging the **user behavior** and using this information, important improvements can be obtained in the **long term**