

# ➤ **Web Information Retrieval**

## **Q&As: Preparation for the exams**

### **SOSE2023**

Frank Hopfgartner, Stefania Zourlidou  
Institute for Web Science and Technologies

- Answer questions posted in padlet
- Answer questions regarding the course, the assignments and the exams
- Discuss a demo exam paper

can you please explain the calculation of rocchio feedback programming question (similar to the assignment) using an example?

1



**zourlidou** 2h

yes, I will do it today in the lecture



Add comment

could you describe the exam pattern? what weightage will be given to each topic that we have covered during weekly assignments?

1



**zourlidou** 2h

I will show a demo test today, so it will become very clear the format of the exam paper. About the weights of the questions: I cannot say, as the exam questions are not yet decided. They will be certainly many topic categories to cover a wide range of topics discussed in the lectures, so the weights will be shaped accordingly.



Add comment

Hi, are we expecting any programming questions in the paper? Also will there be any demo test? Will the question patterns be like the assignments?

3



**zourlidou** 2h

Yes, I will show a demo test in the lecture session today. Programming questions may be given, but they will be a matter of writing a few lines of code. And yes, the questions pattern or format will be very similar with that of assignments.



**Anonymous** 1h

Could you give a general example of how a Programming question could look like in the exam? Or would it be very similar to the questions in the Assignments?



**zourlidou** 1m

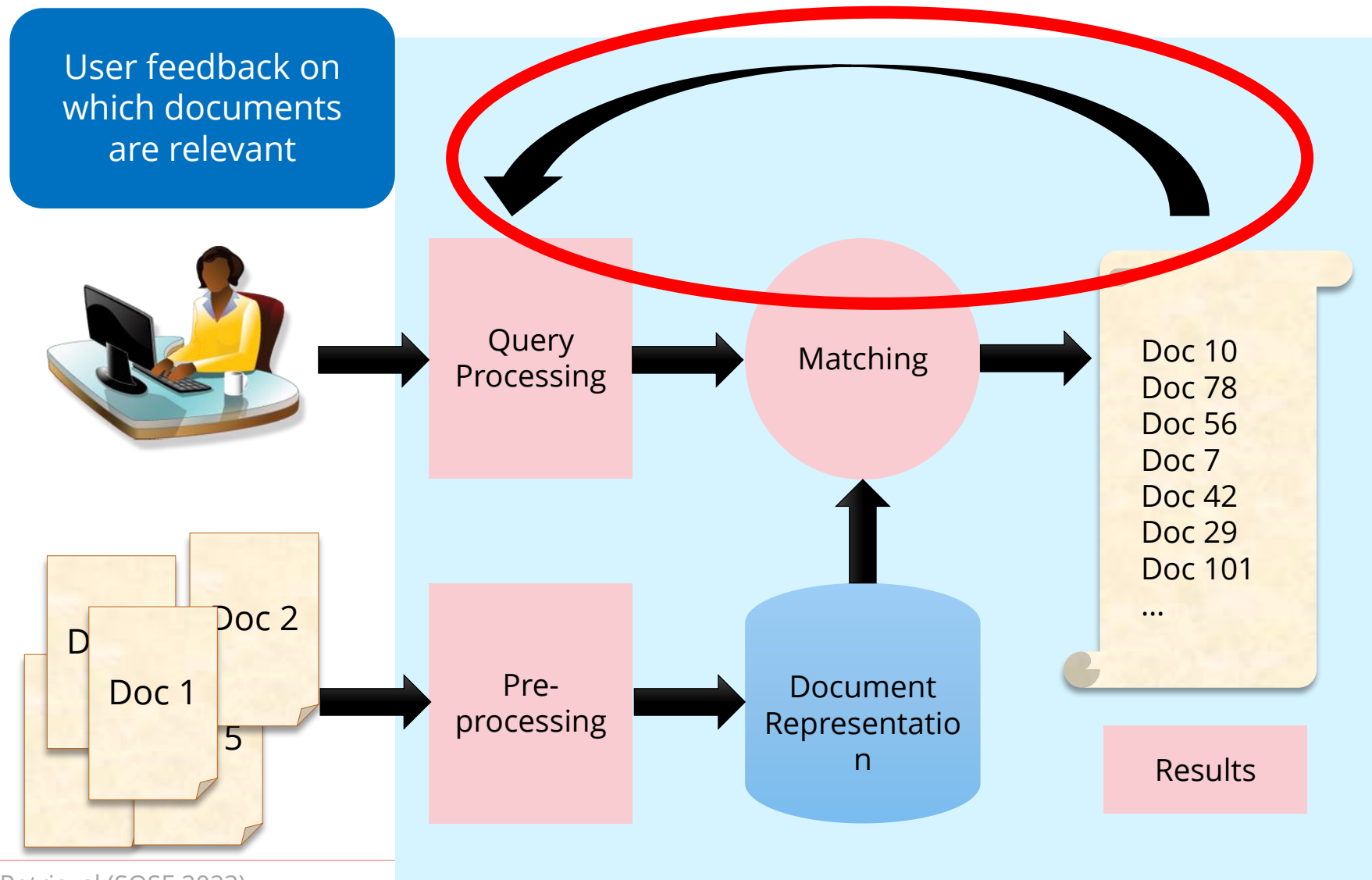
It can be write a function in python, or code will be given where there are some gaps that must be filled in from a given list (I will show a demo today at the lecture at 14:00)



Add comment

## **3. Relevance Feedback (Rocchio)**

# Relevance Feedback for VSM



- User provides for some documents if they are relevant or irrelevant:
  - $D^p$  : set of documents with positive feedback (relevant)
  - $D^n$  : set of documents with negative feedback (irrelevant)
- Adjust query vector

$$\vec{q}_{FB} = \alpha \cdot \vec{q} + \beta \frac{1}{|D^p|} \sum_{d_i \in D^p} \vec{d}_i - \gamma \frac{1}{|D^n|} \sum_{d_i \in D^n} \vec{d}_i$$



- Parameters: typically  $\alpha > \beta > \gamma$
- Adjust negative term weights to 0

# Example

- Query:
  - paris hilton
- Documents

	paris	hilton	hotel	france	eiffel	blonde	heiress	actress
d <sub>1</sub>	3	1	1	2	1			
d <sub>2</sub>	1	3	4	1	3			
d <sub>3</sub>	2	1		1				
d <sub>4</sub>	3			2			1	
d <sub>5</sub>		1	3					
d <sub>6</sub>	3	3				2		
d <sub>7</sub>	2	2					2	
d <sub>8</sub>	2	1	1			1	1	
d <sub>9</sub>	3	2				1		4
d <sub>10</sub>	3	2		1			2	3

# Example

- Initial result list based on VSM
- User provides relevance feedback
  - Positive 
  - Negative 
- Parameter:
  - $\alpha = 1$
  - $\beta = 0.75$
  - $\gamma = 0.15$

Rank	Doc	$\rho$	
1	d <sub>3</sub>	0.395	
2	d <sub>6</sub>	0.183	
3	d <sub>7</sub>	0.161	
4	d <sub>4</sub>	0.132	
5	d <sub>1</sub>	0.128	
6	d <sub>8</sub>	0.125	
7	d <sub>10</sub>	0.071	
8	d <sub>9</sub>	0.057	
9	d <sub>2</sub>	0.049	
10	d <sub>5</sub>	0.027	



# Example

- Adjust query

$$\vec{q}_{FB} = \alpha \cdot \vec{q} + \beta \frac{1}{|D^p|} \sum_{d_i \in D^p} \vec{d}_i - \gamma \frac{1}{|D^n|} \sum_{d_i \in D^n} \vec{d}_i$$

$$\begin{aligned}\alpha &= 1 \\ \beta &= 0.75 \\ \gamma &= 0.15\end{aligned}$$

$$\begin{aligned}\vec{q}_{FB} &= 1 \cdot \begin{pmatrix} 0.046 \\ 0.046 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} + 0.75 \cdot \frac{1}{2} \cdot \begin{pmatrix} 0.137 \\ 0.046 \\ 0.398 \\ 0.602 \\ 0.699 \\ 0 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 0.092 \\ 0.092 \\ 0 \\ 0.301 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \\ &\quad - 0.15 \cdot \frac{1}{2} \cdot \begin{pmatrix} 0.092 \\ 0.092 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0.796 \\ 0 \end{pmatrix} + \begin{pmatrix} 0.092 \\ 0.046 \\ 0 \\ 0 \\ 0 \\ 0.523 \\ 0.398 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} 0.118 \\ 0.087 \\ 0.119 \\ 0.339 \\ 0.262 \\ 0 \\ 0 \\ 0 \end{pmatrix}\end{aligned}$$

Rank	Doc	$\rho$
1	d <sub>1</sub>	0.957
2	d <sub>3</sub>	0.787
3	d <sub>2</sub>	0.691
4	d <sub>4</sub>	0.640
5	d <sub>5</sub>	0.262
6	d <sub>8</sub>	0.172
7	d <sub>10</sub>	0.119
8	d <sub>6</sub>	0.056
9	d <sub>7</sub>	0.050
10	d <sub>9</sub>	0.018



# Other questions?

# Demo exam paper

- [1] <https://olat.vcrp.de/auth/RepositoryEntry/4071063853>
- [2] <https://nlp.stanford.edu/IR-book/information-retrieval-book.html> C. D. Manning, P. Raghavan and H. Schütze, Introduction to Information Retrieval, Cambridge University Press. 2008.
- [3] S. Ceri, A. Bozzon, M. Brambilla, E. Fraternali, S. Quarteroni, Web Information Retrieval, Springer-Verlag Berlin Heidelberg, 2013.
  - Chapter 13, Multimedia Search, available in OLAT.