

# Towards modeling implicit feedback with quantum entanglement

Massimo Melucci

Talk by Emanuele Di Buccio

Department of Information Engineering  
University of Padua



Quantum Interaction  
Oxford, UK, 26th – 28th March 2008

- 1 Information Retrieval
- 2 Towards a Model of Implicit Feedback
- 3 Decomposition
- 4 A Methodology for Implicit Feedback
- 5 Concluding Remarks

# Information Retrieval

- Concerned with retrieving all and only the documents relevant to any information need of any user
- Intrinsically dependent on
  - the user
  - the task the user is performing
- Difficulty of obtaining queries well designed for retrieval purpose
  - vague information need
  - limited experience in the search environment

## Relevance Feedback

- Relevance information explicitly provided by the user used to
  - suggest query expansion terms
  - retrieve new search results
  - dynamically reorder existing results
- Effective but:
  - high user effort
  - benefit not always apparent
- How to remove the burdens of traditional RF?

# Implicit Relevance Feedback

## Alternative: **Implicit Relevance Feedback** (IRF)

- Using contextual features generated during user-information interaction

Example of contextual features: display time, bookmarking, saving, printing, scrolling

- Requires a software tool which
  - monitors subject behavior
  - uses these interaction data as a source of IRF

## Modeling Implicit Feedback

- Abstraction of the software tool to automate the feedback process
- Geometric Framework proposed in [1] to model Implicit Feedback
  - considers multiple source of evidence
  - personalized for each user
  - tailored for each search task
- Framework revisited with some QM concepts according to the proposal originally reported in [2] where QM was proposed for tackling the problem of IR

- [1] M. Melucci and R. W. White.  
Utilizing a geometry of context for enhanced implicit feedback.  
In *Proceedings of the ACM Conference on Information and Knowledge Management (CIKM)*, pp. 273–282, Lisbon, Portugal, November 2007.
- [2] C.J. van Rijsbergen.  
*The Geometry of Information Retrieval*.  
Cambridge University Press, New York, 2004.

# Towards a Model of Implicit Feedback

# Towards a Model of Implicit Feedback

User-document interaction

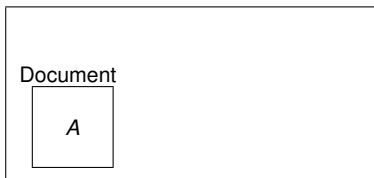


Information Retrieval



# Towards a Model of Implicit Feedback

User-document interaction

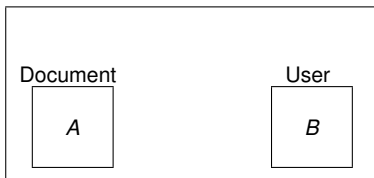


## Information Retrieval

**A** visited document or the visit of the document

# Towards a Model of Implicit Feedback

User-document interaction

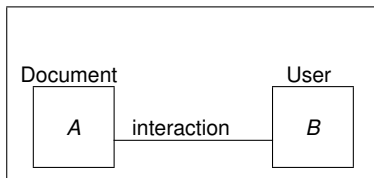


## Information Retrieval

- A visited document or the visit of the document
- B user, i.e. assessment the user provides about the document

# Towards a Model of Implicit Feedback

## User-document interaction

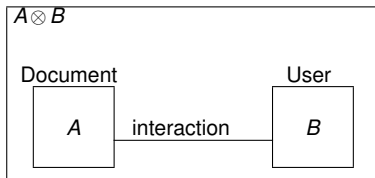


## Information Retrieval

- A** visited document or the visit of the document
- B** user, i.e. assessment the user provides about the document
- AB** user–document interaction as a composite system

# Towards a Model of Implicit Feedback

## User-document interaction

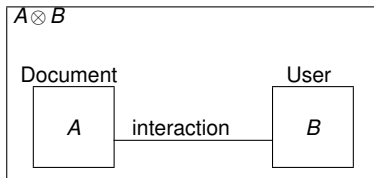


## Information Retrieval

- A** visited document or the visit of the document
- B** user, i.e. assessment the user provides about the document
- AB** user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

# Towards a Model of Implicit Feedback

User-document interaction



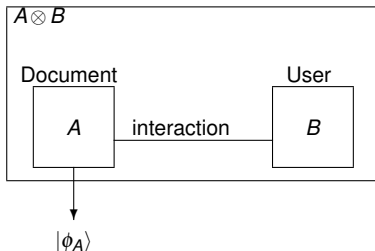
## Information Retrieval

## Quantum Mechanics

- A** visited document or the visit of the document
- B** user, i.e. assessment the user provides about the document
- AB** user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

# Towards a Model of Implicit Feedback

User-document interaction



## Information Retrieval

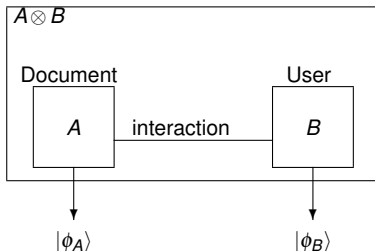
- $A$  visited document or the visit of the document
- $B$  user, i.e. assessment the user provides about the document
- $AB$  user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

## Quantum Mechanics

- $|\phi_A\rangle$  abstraction of the visited document – reduced to a vector

## Towards a Model of Implicit Feedback

User-document interaction



### Information Retrieval

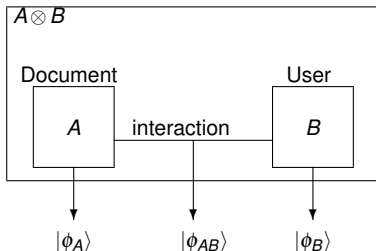
- A visited document or the visit of the document
- B user, i.e. assessment the user provides about the document
- AB user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

### Quantum Mechanics

- $|\phi_A\rangle$  abstraction of the visited document – reduced to a vector
- $|\phi_B\rangle$  abstraction of the user who visited the document – reduced to a vector

# Towards a Model of Implicit Feedback

## User-document interaction



## Information Retrieval

- A** visited document or the visit of the document
- B** user, i.e. assessment the user provides about the document
- AB** user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

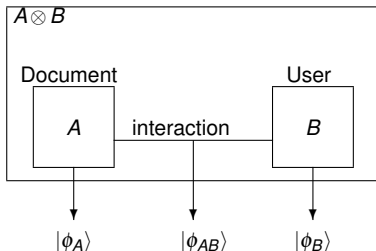
## Quantum Mechanics

- $|\phi_A\rangle$  abstraction of the visited document – reduced to a vector
- $|\phi_B\rangle$  abstraction of the user who visited the document – reduced to a vector
- $|\phi_{AB}\rangle$  interaction between  $A$  and  $B$  element of  $A \otimes B$



# Towards a Model of Implicit Feedback

## User-document interaction



## Information Retrieval

- A** visited document or the visit of the document
- B** user, i.e. assessment the user provides about the document
- AB** user–document interaction as a composite system
  - relationship between  $A$  and  $B$ ?

## Quantum Mechanics

- $|\phi_A\rangle$  abstraction of the visited document – reduced to a vector
- $|\phi_B\rangle$  abstraction of the user who visited the document – reduced to a vector
- $|\phi_{AB}\rangle$  interaction between  $A$  and  $B$  element of  $A \otimes B$ 
  - $|\phi_{AB}\rangle$  entangled or separable?

# Schmidt Decomposition

## Bipartite System $AB$

- Bipartite pure state  $|\phi_{AB}\rangle$  separable iff can be written as

$$|\phi_{AB}\rangle = |\phi_A\rangle \otimes |\phi_B\rangle$$

- In general  $|\phi_{AB}\rangle$  is written in any orthonormal product basis  $\{|i_A\rangle \otimes |j_B\rangle\}$  as

$$|\phi_{AB}\rangle = \sum_{i=0}^{N_A} \sum_{j=0}^{N_B} C_{ij} |i_A\rangle \otimes |j_B\rangle \quad \text{where } C_{ij} \text{ element of a matrix } C$$

- Schmidt Decomposition Theorem

There always exists a product bi-orthonormal basis  $\{|e_A^i\rangle \otimes |e_B^i\rangle\}$

$$|\phi_{AB}\rangle = \sum_i^{\min\{N_A, N_B\}} \lambda_i |e_A^i\rangle \otimes |e_B^i\rangle$$

where  $\lambda_i$  nonzero singular eigenvalues of  $C$ .

$\lambda_i$ : obtained by Singular Value Decomposition of  $C \rightarrow C = B\Lambda A \rightarrow \lambda_i = \{\Lambda\}_{i,i}$

**Schmidt Number**: number of nonzero singular eigenvalues of  $C$

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

## Interaction Data

- subset of the document features of the dataset

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

## Interaction Data

- subset of the document features of the dataset
  - display time
  - bookmark
  - saving
  - access frequency
  - scrolling

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

## Interaction Data

- subset of the document features of the dataset
  - display time
  - bookmark
  - saving
  - access frequency
  - scrolling
- usefulness scores assigned to each document

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement



# A Methodology for Implicit Feedback

normalization of the dataset

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

# A Methodology for Implicit Feedback

normalization of the dataset



choice of a subset of subject IDs and a subset of task IDs

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

normalization of the dataset



choice of a subset of subject IDs and a subset of task IDs



selection of the five features for each subject

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

normalization of the dataset



choice of a subset of subject IDs and a subset of task IDs



selection of the five features for each subject



computation of the average value of every feature from the tuples of the chosen subject and task identifiers

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

normalization of the dataset



choice of a subset of subject IDs and a subset of task IDs



selection of the five features for each subject



computation of the average value of every feature from the tuples of the chosen subject and task identifiers



grouping average values by usefulness score into a binary scale

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 **computation of a contingency matrix**
- 3 decomposition of the contingency matrix
- 4 analysis of entanglement

normalization of the dataset



choice of a subset of subject IDs and a subset of task IDs



selection of the five features for each subject



computation of the average value of every feature from the tuples of the chosen subject and task identifiers



grouping average values by usefulness score into a binary scale



normalization of the obtained  $2 \times 5$  matrix

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 **decomposition of the contingency matrix**
- 4 analysis of entanglement

# A Methodology for Implicit Feedback

## Possible Methodology

- 1 preparation of the interaction data
- 2 computation of a contingency matrix
- 3 decomposition of the contingency matrix
- 4 **analysis of entanglement**



## Example - All subjects and All tasks

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \text{useless} \\ -0.162 & -0.066 & 0.020 & -0.037 & 0.004 & \\ 0.201 & 0.082 & -0.025 & 0.043 & -0.005 & \text{useful} \end{bmatrix}$$

## Example - All subjects and All tasks

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \\ -0.162 & -0.066 & 0.020 & -0.037 & 0.004 & \text{useless} \\ 0.201 & 0.082 & -0.025 & 0.043 & -0.005 & \text{useful} \end{bmatrix}$$

Decomposition of the contingency matrix: SVD of  $\mathbf{C} \Rightarrow \mathbf{C} = \mathbf{B}\Lambda\mathbf{A}$

$$\Lambda = \begin{bmatrix} 0.9999 & 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ 0.0000 & 0.0059 & 0.0000 & 0.0000 & 0.0000 \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} \\ 0.902 & 0.182 & 0.108 & -0.375 & 0.023 \\ 0.367 & 0.074 & 0.033 & 0.927 & 0.007 \\ & & \dots & & \end{bmatrix}$$

## Example - All subjects and All tasks

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \text{useless} \\ -0.162 & -0.066 & 0.020 & -0.037 & 0.004 & \\ 0.201 & 0.082 & -0.025 & 0.043 & -0.005 & \text{useful} \end{bmatrix}$$

$$\Lambda = \begin{bmatrix} 0.9999 & 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ 0.0000 & 0.0059 & 0.0000 & 0.0000 & 0.0000 \end{bmatrix}$$

### Analysis of the Entanglement

- $\lambda_{1,1} = 0.9999 \approx 1 \Rightarrow$  *Schmidt Decomposition*  $|\phi_{AB}\rangle \approx |1_A\rangle \otimes |1_B\rangle$
- all the subjects and tasks:  $A \otimes B$  is a product state vector
- is entanglement somehow related to the subject or to the task?

## Example - Subject 1 and Task 1

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \\ -0.158 & -0.094 & -0.052 & -0.110 & 0.835 & \text{useless} \\ 0.014 & 0.128 & -0.052 & 0.256 & 0.411 & \text{useful} \end{bmatrix}$$

## Example - Subject 1 and Task 1

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \\ -0.158 & -0.094 & -0.052 & -0.110 & 0.835 & \text{useless} \\ 0.014 & 0.128 & -0.052 & 0.256 & 0.411 & \text{useful} \end{bmatrix}$$

Decomposition of the contingency matrix: SVD of  $\mathbf{C} \Rightarrow \mathbf{C} = \mathbf{B}\mathbf{\Lambda}\mathbf{A}$

$$\mathbf{\Lambda} = \begin{bmatrix} 0.94 & 0.00 & 0.00 & 0.00 & 0.00 \\ 0.00 & 0.33 & 0.00 & 0.00 & 0.00 \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} \\ 0.145 & 0.243 & -0.053 & -0.188 & 0.939 \\ 0.031 & 0.473 & 0.092 & -0.827 & -0.288 \\ & & \dots & & \end{bmatrix}$$

## Example - Subject 1 and Task 1

$$\mathbf{C} = \begin{bmatrix} \text{display time} & \text{bookmark} & \text{saving} & \text{frequency} & \text{scrolling} & \\ -0.158 & -0.094 & -0.052 & -0.110 & 0.835 & \text{useless} \\ 0.014 & 0.128 & -0.052 & 0.256 & 0.411 & \text{useful} \end{bmatrix}$$

$$\Lambda = \begin{bmatrix} 0.94 & 0.00 & 0.00 & 0.00 & 0.00 \\ 0.00 & 0.33 & 0.00 & 0.00 & 0.00 \end{bmatrix}$$

### Analysis of the Entanglement

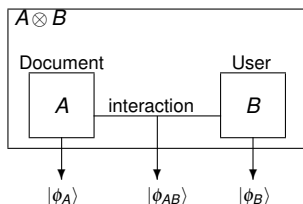
Second singular value is significantly greater than 0

⇒ Schmidt number is significantly greater than 1

⇒  $|\phi_{AB}\rangle$  is entangled

## Concluding Remarks and Open Issues

User-document interaction



- User visiting a document
  - ⇒ state of the system (A) is the pure state of the observable which describes the measurement of the style of interaction
- User assessing the usefulness of a document
  - ⇒ the state of the system (B) is to the pure state of the observable that describes the measurement of usefulness
- State of  $A \otimes B$  separable
  - ⇒ measurement of a property does not influence the state of the other observable
- What happens if the composite state of  $A \otimes B$  is entangled?

# Open Issues

- When the composite state of  $A \otimes B$  is entangled
  - how does the state of B behave?
  - what pure state of B does it collapse to?
- When does the collapse occur, given that it occurs?
- Entanglement used to verify the existence of a relationship between user and information.

How to use it for retrieval purpose?