

### First VITALAS Dissemination Conference D8.2.1

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### **ABSTRACT:**

This document contains proceedings of the 1st VITALAS Dissemination Conference, held in Cannes, France, on 11-12 February 2008, jointly with the NoE MUSCLE Final Conference. The conference clustered members of communities in the field of Multimedia Understanding through Semantics, Computation and Machine Learning.

The main objective of this conference was to present achieved results within the VITALAS project

KEYWORD LIST: Dissemination, promotion visibility, conference

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- Daniel Schneider, FhG
- Gérard Dupont, EADS
- Hervé Goëau, INA

### List of Evaluators

- Joost Geurts, INRIA

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### 1. Introduction



The MCVC'08 (MUSCLE Conference joint with VITALAS Conference 2008) held in Cannes, France, on 11 & 12 February 2008, clustered members of communities in the field of Multimedia Understanding through Semantics, Computation and Machine Learning.

This scientific conference was co-chaired by Nozha Boujemaa (scientific coordinator of MUSCLE and VITALAS) and Eric Pauwels (CWI)

In addition to the keynote talks, this event included MUSCLE e-Team presentations and showcase demonstrations as well as presentations from other EU-funded projects: VITALAS and PASCAL.

For each VITALAS presentation, a brief description is reported in section3, and slides are set in section 5.

### 2. Agenda and list of participants

### 2.1 Agenda

### **Monday 11 February 2008**

- **09:40 10:00** Registration Welcome coffee
- 10:00 10:15 Introduction MUSCLE achievements, Nozha Boujemaa, INRIA Imedia
- 10:15 11:00 Keynote speaker: Prof. Andrew Zisserman, University of Oxford
- 11:00 11:30 Coffee break Muscle showcase demonstrations
- **11:30 12:00** Shape modelling via higher-order active contours and phase fields *Ian Jermyn*, INRIA Ariana
- 12:00 12:30 A contrario matching of local features between images, Yann Gousseau, ENST
- 12:30 14:00 Lunch
- **14:00 14:30** Recognising Animals, *Allan Hanbury*, TU Vienna-PRIP

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- 14:30 15:00 Person detection and recognition, tracking and analysis Montse pardas, UPC
- 15:00 15:30 Action class detection and recognition Ivan Laptev, INRIA Vista
- 15:30 16:00 Coffee break MUSCLE showcase demonstrations
- **16:00 16:30** On Sequence Kernels for SVM classification of sets of vector *Khalid Daoudi*, IRIT
- **16:30 17:00** Active machine learning based on kernels on bags and on graphs *Sylvie Philipp-Foliguet*, ENSEA
- 17:00 17:20 Crossing textual and visual content in different application scenarios *Gabriela Csurka & Jean-Michel Renders*, Xerox (PASCAL)
- 17:20 17:50 Fully Bayesian Source Separation with Application to the CMB Simon Wilson, TCD

### **Tuesday 12 February 2008**

- **09:00 09:45** Keynote speaker: *Joachim Köhler*, Fraunhofer IAIS, "Audio Content Search"
- 09:45 10:15 Audio-Visual Speech Analysis & Recognition Nassos Katsamanis, NTUA
- 10:15 10:45 Multimodal Interfaces Alexandros Potamianos, TUC
- **10:45 11:15** Coffee break <u>MUSCLE showcase</u> demonstrations
- **11:15 11:45** Multimodal Processing and Multimedia Understanding: Image Retrieval Using Eye Movements *Fred Stentiford*, UCL
- 11:45 12:05 Interactive Visualization tool with Graphic Table of Video Contents *Hervé Goëau*, INA (VITALAS)
- **12:05 12:30** Open Vocabulary Speech Analysis in VITALAS, *Daniel Schneider*, FhG (VITALAS)
- 12:30 14:00 Lunch
- **14:00 14:30** Feature extraction from audio and their application in music organization and transient enhancement in recorded music, *Jakob Frank*, TUWIEN and *Massimo Magrini*, CNR-ISTI
- **14:30 15:00** BilVideo: MPEG-7 Compliant Video Database Management System, *Onur Kucuktunc*, Bilkent University
- **15:00 15:20** Implicit feedback learning in semantic and collaborative information retrieval systems, *Gérard Dupont*, EADS (VITALAS)
- **15:30** End

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### 2.2 List of participants

Amiaz Tomer Tel Aviv University

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Toreyin B.Ugur Bilkent University
Wilson Simon Trinity College Dublin
Zisserman Andrew University of Oxford

### 3. VITALAS presentations

This section gives the abstracts from the presentations provided by the presenters. The slides are annexed at the end of this document.

### Interactive visualization tool with graphic table of video contents

Hervé Goëau INA

In this paper, we present an interactive visualization, called Table Of Video Contents (TOVC), for browsing structured TV programs such as news, magazines or sports. In these telecasts, getting a good segmentation can be very time-consuming, especially in an annotating context. Our visualization, connected with a classical media player, offers a very handy video browser. This system allows a global overview by showing the temporal structure and by giving some semantic information. The drawn structure enables a non linear video access by suggesting relevant key frames. The TOVC is created from a graphic framework designed for computing similarities on visual contents, and displaying the associated proximities in a 2D map with graph representation. TOVC is one of its first applications and shows interesting capabilities.

### Open Vocabulary Speech Analysis in VITALAS

Daniel Schneider FhG IAIS

Automatic indexing of TV and radio speech data requires robust components for both speech recognition and spoken document retrieval. Due to the high topic variability and the resulting large vocabularies, classic word-based approaches have to cope with a high number of out-of-vocabulary words. This talk presents a phonetic approach to open vocabulary indexing based on syllable decoding and retrieval. Current experimental results are presented,

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followed by a demonstration of the Fraunhofer IAIS AudioMining system for spoken term detection.

### Implicit feedback learning in semantic and collaborative information retrieval systems

Gérard Dupont EADS

Information retrieval is a very wide domain which can involve various types of activities and tasks. Many complex factors are participating in a search for information and many systems have been experimented. Nowadays a general consensus has been established around a keyword/document matching process which appears to be efficient on large scale and have enough reliability to satisfy a significant part of the users. Btu this claim has to be limited and for some subjects, search is still a difficult task. Many reasons can be proposed to explain this phenomenon, but the most salient ones are the difficulty for users to express their needs while searching for information and the limitation of shared knowledge between users and information retrieval systems, meaning that both users and machines don't really understand the information and knowledge space used as references by the other.

This presentation try to provide an overview of one way to resolve those gaps: using feedback learning. The aim is to make the system learning on user behaviour in order to better define its current needs. Machine learning algorithms applied on signal coming from user while performing a search can lead to the understanding of what is really relevant to the users and then can be exploited to help him during its tasks.

The work, engaged through the VITALAS1 project, is presented: study of users search logs and definition of a feedback learning framework. Then research on implicit relevance feedback and query optimisation is presented as a first attempt to exploit the feedback learning framework. Finally an overview of the next steps within those studies is presented and especially their impact on the VITALAS project.

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### 4. Dissemination impact

This Conference was open to the scientific community in the field of Multimedia Understanding through Semantics, Computation and Machine Learning.

In order to relay the event, the conference announcement was posted on the VITALAS project website and on the website of CHORUS, the European Coordination Action. After the event, video recordings of the presentations (recorded by videolectures.net) were made available through the project website.



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The Conference is open to the scientific community at large, including the whole Muscle Community, the European Commission, as well as all EU funded projects. This event will be an opportunity for MUSCLE NoE to highlight and demonstrate to all, the major results achieved by the members of its community in the field of Multimedia Understanding through Semantics, Computation and Machine Learning. The MUSCLE scientific conference will be co-chaired by Nozha Boujemaa and Eric Pauwels In addition to the keynote talks, this event includes MUSCLE e-Team presentations and showcase demonstrations as well as presentations from other EU-funded projects.

### Introduction



### Lectures



Automated Character Annotation in Multimedia Andrew Zisserman



Shape modelling via higher-order active contours and phase fields Ian Jermyn



A contrario matching of local features between images Yann Gousseau



Recognising Animals Allan Hanbury









25 April 2008 Page 11 of 29 ERCIM took the opportunity of this 1st conference to improve the VITALAS poster, assigning a communication agency. The result is of a great quality and the 60x80cm posters have been distributed to each partner for dissemination at a local level.



### 5. Conclusion

International researchers from different scientific disciplines participated in the VITALAS Conference and engaged interesting and fruitful discussions with VITALAS project partners. Based on positive feedback from the participants and the number of visitors who viewed the video presentations after the event, we may consider the VITALAS conference a success. The next VITALAS Conference is planned for February 2009.

### 6. Annex: Presentations

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# INTERACTIVE VISUALIZATION TOOL WITH GRAPHIC TABLE OF VIDEO CONTENTS







Hervé Goëau

## INTERACTIVE VISUALIZATION TOOL WITH GRAPHIC TABLE OF VIDEO CONTENTS

- 1. Context
- Solutions in the state of the art
- 3. Framework and method
- 4. Results
- Evaluation
- 3. Future works

### $\sim$

## **Context and objectives**

- 51 channels daily collected (100 in 2010)
- **Constant** staff of **80** archivists (DL)
- News telecast : a strategic issue
- very time consuming
- one daily program for each main channel
- finest level of annotation
- →accelerate the access to the content by giving a quick overview

Basic

Time line or grid

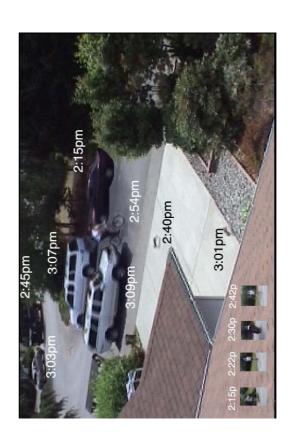


video slider



Lack of scalability

Merging



one shot, no global motion

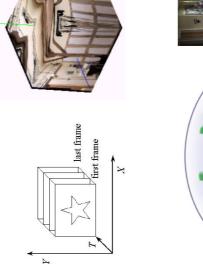
Stiching - dynamic

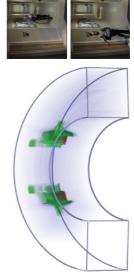


page setting

### 3D representation







Readability?

### $\infty$

### Video summarization and browsing interfaces

### Keyframes tiling







Semantic analysis of content?

### Our solution

Table Of Video Contents (TOVC)

Highlights the video structure

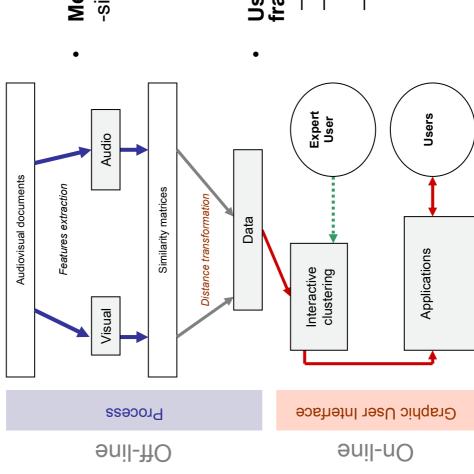
→ non linear access

→ fast access

→ location in the content

### 10

## **Generic Framework**



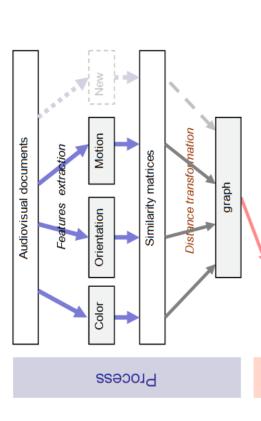
Media analysis:

-similarity measures computation

### User oriented experimental framework:

- video browsing with TOVC
- semi-supervised or unsupervised photo annotation
- navigation, proximity exploration

# Current implementation of TOVC



- Media analysis:
- visual descriptors extraction
- similarity measures computation
- MultiDimensional Scaling (MDS)
- Graph representation
- numerous drawing algorithms

EXPERT USER

Fusion, filtering

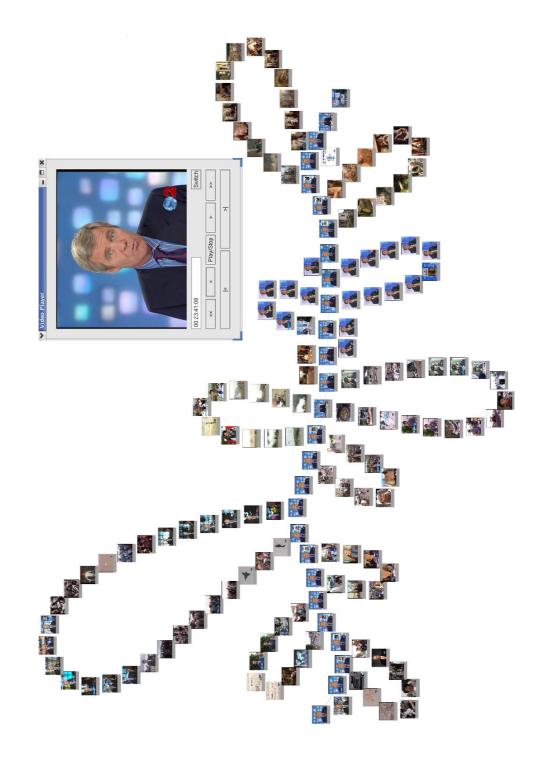
Clustering

Graphic User Interface

Graph layout

USERS

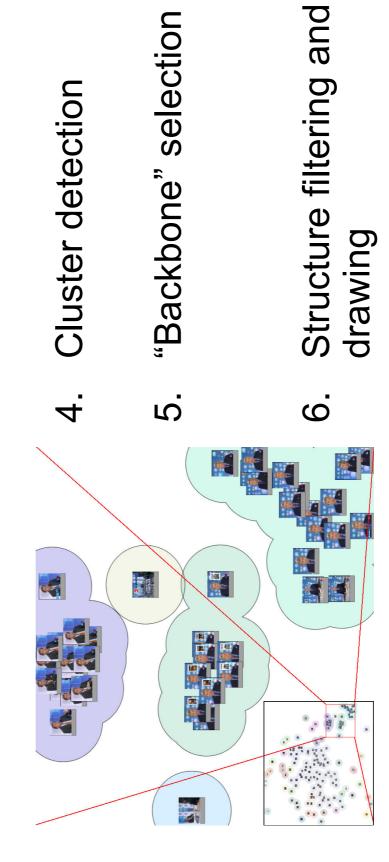
Video browsing application

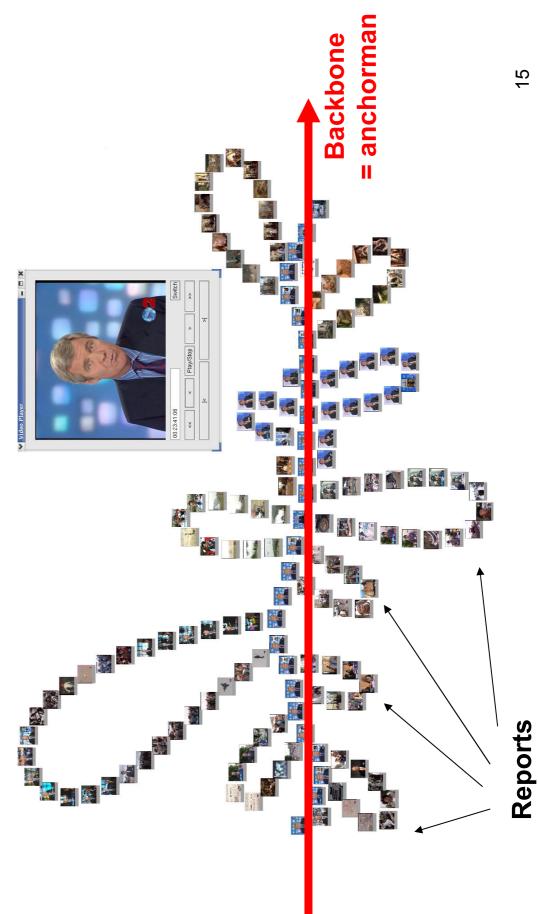


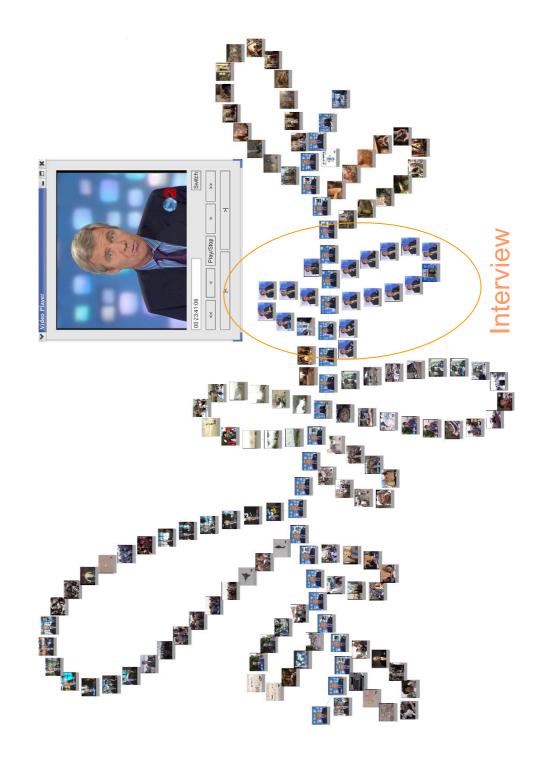
 No key frame detection (constant frequency extraction)

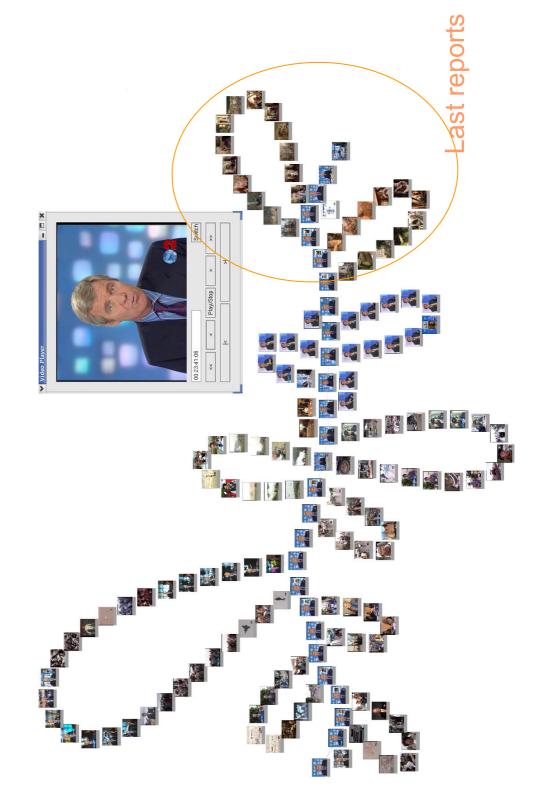
2. Feature combination

Instantaneous feedback by using MDS





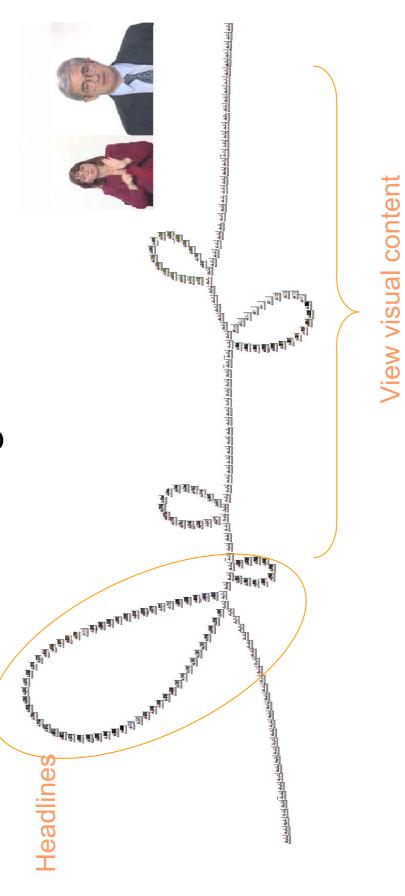




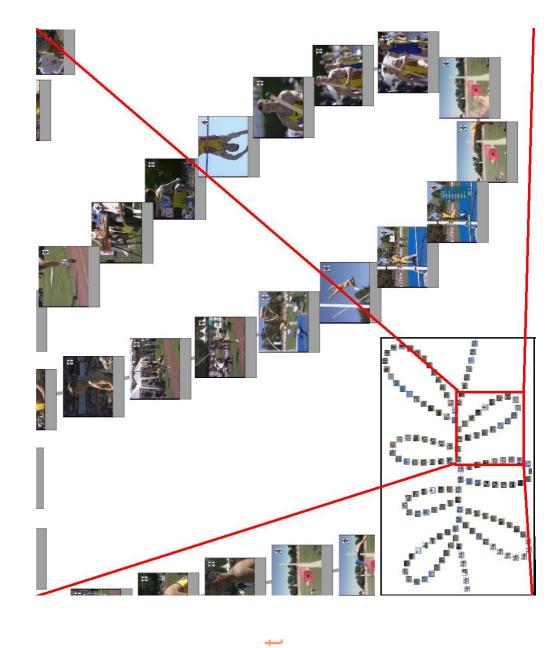
### 19

### Some Results in Various **Programs**

deaf and hard hearing or news in brief



## TOVC in sport program



One loop = = one event

# Evaluation / usability test

### How to evaluate a user interface?

- 12 users
- **TOVC versus basic** grid visualization
- 2 news programs
- Timed search tasks
- Interview

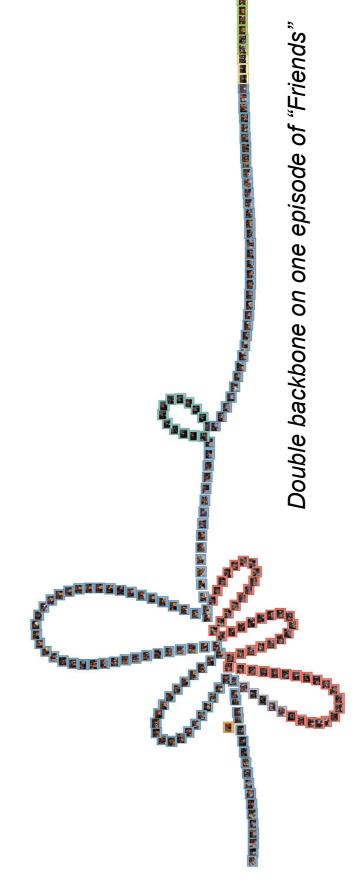


## **Evaluation Results**

- Search task :
- 50% are faster with TOVC
- 25% indifferent
- 25% are more efficient with the grid view
- 20% time average on search task
- Interview:
- localization in the video content
- remember an already seen frame

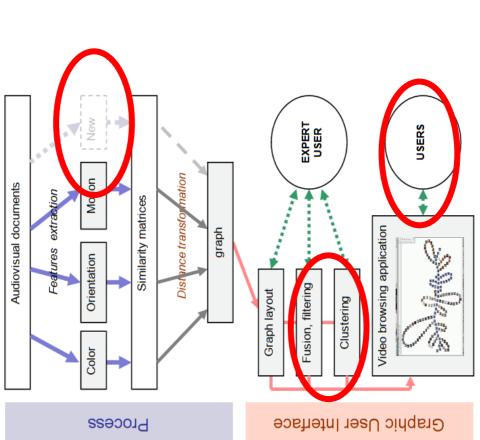
## **Future works**

Experiment on less structured video like fictions, series, documentaries



## **Future works**

Audio description



Fusion formalism

Learning process (for news)

evaluation process

#### **Open Vocabulary Speech Analysis in Vitalas**



Daniel Schneider Speech Group, Fraunhofer IAIS



#### **Outline**

- Vitalas Scenario: Broadcast News Audio Indexing
- Structural Audio Analysis
- Open Vocabulary Speech Recognition
- Demo: AudioMining

#### Challenge in Vitalas: Large Scale Broadcast News Indexing

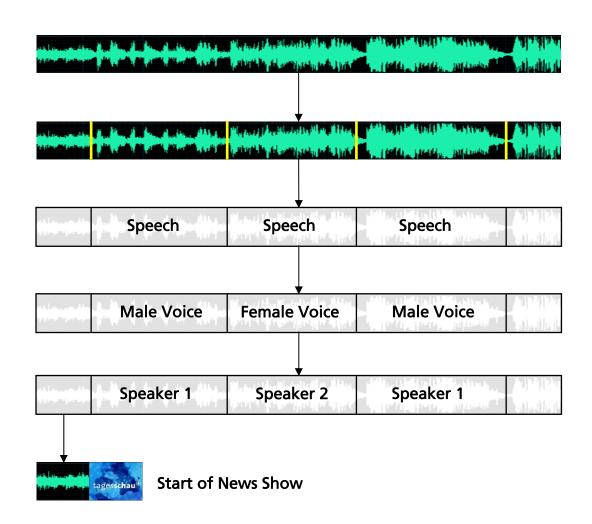
- Huge amount of data (> 10.000 hours)
- Heterogeneous material
  - From various sources of unknown type
  - High topic variability
  - Huge vocabulary
  - Multilingual data
- Requires efficient and robust algorithms for...
  - Information extraction
  - Information retrieval





#### **Structural Audio Analysis in Vitalas**

- Unstructured Audio Data
- Homogeneous Segmentation
- Speech Detection
- Gender Detection
- Speaker Clustering
- Programme Identification via Jingle





**Transcript 3** 

#### **Speech Recognition**

**Speech Recognition** 

**Transcript 1** 

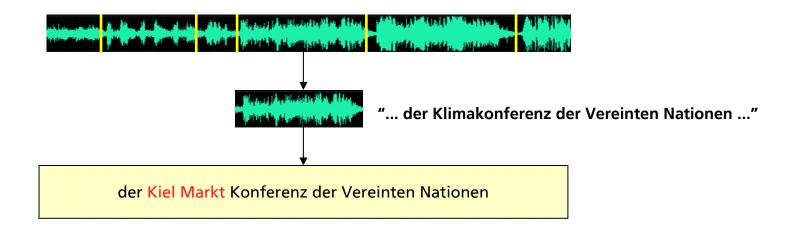
**Transcript 2** 

- Transcripts can be used for...
  - Search in entire archive ("Audio-Google")
  - Media observation (Alert if keyword occurs)
  - Input for text mining (e.g. Topic Detection)



#### **Speech Recognition Challenges**

- Out-of-Vocabulary problems with classic word based ASR of broadcast data
  - New and popular words (e.g. Gammelfleischskandal "rotten meat scandal")
  - Proper names (companies, cities, people)
- Compound words in German (climate conference Klimakonferenz)
- Huge Lexica required large effort



#### Phonetic Approach to Open Vocabulary Indexing

- Idea:
  - Search on phonetic subword level instead of word level
  - Search for a sequence of sounds instead of words



#### Phonetic Approach to Open Vocabulary Indexing

1. Generate transcription on subword level (phone or syllable)



2. Break down search term into subword units

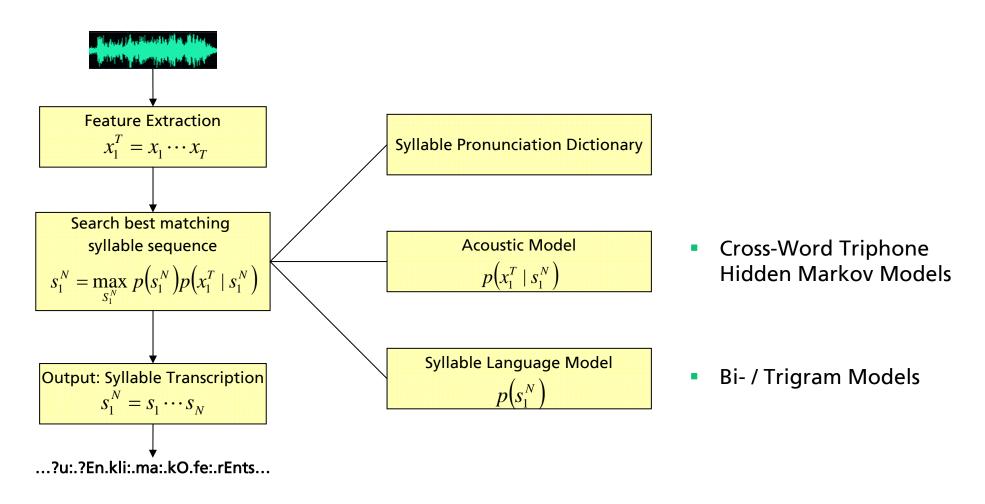
Klimakonferenz → kli:.ma:.kOn.fe:.rEnts

3. Fuzzy Phonetic Search





#### Phonetic Approach (1): Generate Subword Transcription



#### Phonetic Approach (2): Fuzzy Syllable Search

- Break down search term: Klimakonferenz → kli:.ma:.kOn.fe:.rEnts
- Goal: Retrieval of documents containing similar syllable sequences
- Fuzzy search based on Levenshtein Distance between
  - Single syllables
  - Syllable sequences

•	Examples distances between single syllables:				
d_e:_s_	d_e:_s_	zero			
d_e:_s_	k_O_n_	high			
d_e:_s_	d_i:_s_	low			
d_e:_s_	d_l_s_	medium			

Examples distances between syllable				
sequence:				
k_l_i:_ m_a:_	k_l_i:_ m_a:_	zero		
k_l_i:_ m_a:_	k_l_i:_ n_a:_	low		
k_l_i:_ m_a:_	k_l_i:_ n_6_	high		

Solution based on Dynamic Programming (c.f. Speech Decoding)

#### **Properties of Phonetic Subword Approach**

- The set of subword units is finite and (rather) small
  - Complete vocabulary coverage (no OOV)
  - 10.000 syllables compared to 300.000+ words
  - Compact ASR search space
- Implicit decomposition of compounded words
  - kli:.ma:.kOn.fe:.rEnts gives 100% hit for the search terms Klima, Konferenz, Klima Konferenz, Klimakonfernez
- Implicit stemming capabilities of fuzzy search
  - Skandal skan.da:l
     Skandals skan.da:ls (less important to learn genitive explicitly)



#### **Experiments: Fraunhofer AudioMining Corpus**

- High Quality Studio Data
  - Accurate sentence level transcriptions
  - (Almost) no background noise
  - Only one speaker per segment
- 14 hours of carefully annotated training data
- 3 hours of evaluation data (disjoint from training set)

- Main Challenges
  - Speaking rate (interview vs. read speech)
  - Spontaneous Speech in interview situation

**Data: German News Shows** 

Comparable to VITALAS

data sets from IRT and INA







#### **Experiments: Model Setup**

- Acoustic Models
  - Maximum Likelihood Reestimation
  - Phonetic Clustering of triphones
  - 7300 triphone HMMs with up to 16 Gaussian mixture components
- Language Models trained on 2000-2006 newswire data with CMU SLM toolkit
  - 80 million running words
  - Text transformed to syllables
  - Corpus Topics: Politics, Economy, Culture, Sports
- Pronunciation Lexicon: 10000 most frequent syllables from LM training



#### **Current Results - Speech Recognition**

Task: Syllable Transcription of 3 hours of Broadcast Data (Radio Shows)

Syllable Error Rate	ASR Realtime Factor	
34.3	1.5	

- High error rates (test set includes several BC shows)
- Example for frequent substitution error:
  - Reference: U\_n\_t\_ (and)
  - Recognized: U\_n\_ (an')
- Errors partly covered by fuzzy retrieval

#### **Current Results – Fuzzy Phonetic Retrieval**

- Task: Detect 213 keywords and keyphrases in recognition results
- Confidence thresholds of the fuzzy search can be chosen depending on the application

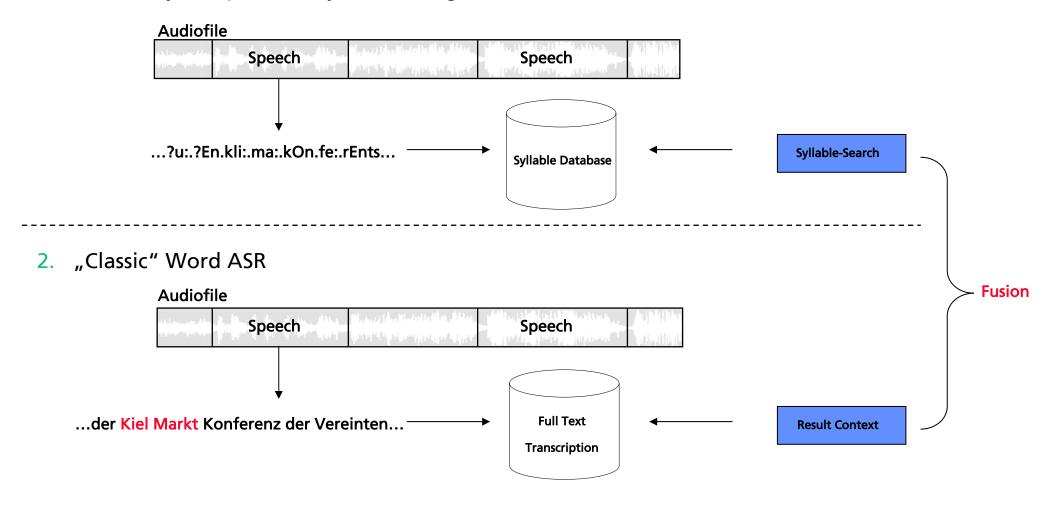
Confidence Threshold	Precision	Recall	Remark
0.70	0.66	0.65	Equal Error
0.85	0.91	0.53	Tuned for Precision

- Some errors due to...
  - Search term is substring of actual spoken compound word (Klima Klimakonferenz)
  - Short search terms consisting of highly frequent syllables (Mutter mU.t6)



#### **Additional Word Context for Enhanced Display of Results**

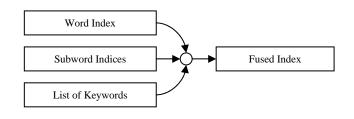
1. Vocabulary Independent Syllable Recognition



#### **Demo: AudioMining**

#### **Next Steps**

- Evaluate the syllable approach on other languages
  - Vitalas End-Users: IRT (German) and INA (French)
- Improve Recognition Accuracy
  - Use information extracted by structural analysis
  - Speaker / domain / progamme adaptivity
- Improve Information Retrieval Accuracy
  - Fusion of word, syllable and phoneme recognition results
  - Exploit ASR output graph instead of 1-Best



- Consider Scalability
  - Current search approach not applicable for 10k hours archive
  - Evaluate efficient implementations and alternatives



#### Thank you!



### Implicit feedback learning in semantic and collaborative information retrieval systems



by Gérard Dupont<sup>1, 2</sup>

written under the direction of

Sébastien Adam<sup>1</sup>, Yves Lecourtier<sup>1</sup>, Bruno Grilheres<sup>1</sup>, <sup>2</sup>, Stephan Brunessaux<sup>2</sup>

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- <sup>2</sup> **EADS Defense and Security**, Information Processing and Competence Center Val de Reuil, France

#### Summary

- Introduction
- Enhanced IRS with feedback learning
- Feedback learning in VITALAS
- Focus on learning using behavior measure as feedback
- Conclusion and future work





#### Introduction



#### **Information retrieval?**



"Information retrieval (IR) is finding material (usually documents) of an unstructured nature (usually text) that satisfy an information need from within large collections (usually stored on computers)"

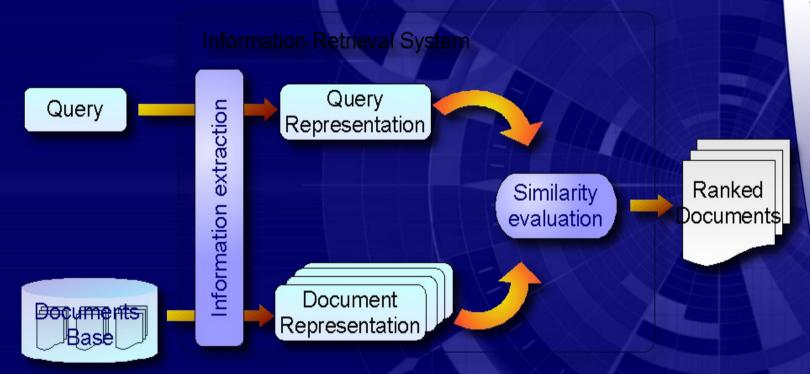
An introduction to information retrieval - Manning, 2007.

#### Variables:

- —Document and collection of documents: library, database, Intranet, Internet...
- -Unstructured information without precise meaning
- Information needs expressed by users

### EADS DEFENCE & SECURITY

#### Simple view of IRS



Aim: Matching query with documents (or part of documents)

- Information model to represent document and needs
- Similarity evaluation theory to produce ranked list of documents

#### Information model

Vector model dedicated to text document retrieval



#### Document term vectors

$$ec{D}_i = egin{array}{c} word_1 - 0.24 \ word_2 - 0.4 \ word_3 - 0.1 \ ... \ word_N - 0.1 \ \end{array}$$

Query term vectors

$$ec{Q} = egin{array}{c} word_1 - 0 \ word_2 - 1 \ word_3 - 0 \ ... \ word_N - 1 \ \end{array}$$

Example of similarity formula (normalized cosinus)

$$score(\vec{D}_i, \vec{Q}) = \frac{\langle \vec{D}_i, \vec{Q} \rangle}{\|\vec{D}_i\|.\|\vec{Q}\|} = \frac{\sum_{k=1}^{N} d_{i,k} \times q_k}{\sqrt{\sum_{k=1}^{N} d_{i,k}^2} \times \sqrt{\sum_{k=1}^{N} q_k^2}}$$

Generalized probabilistic model

$$P(D|L) = \prod_{i} P(A_{i} = a_{i}|L)$$

- Term vector extended to description through attributes/values
- Relevance as probability
- Possibility to handle multimedia features as attributes

#### **Limits of current IRS**

## EADS DEFENCE & SECURITY V TALAS

#### Strong assumptions:

- Dimensions of the vector model or attributes in the probabilistic model are independent
- User information needs is fully described by its query

#### Not verified in most of the cases:

- Linguistic study tells us that terms are not independent in texts
   (synonymous, antonymous,...) neither are features extracted in CBIR
- User can not define precisely their needs since they are trying to complete their knowledge

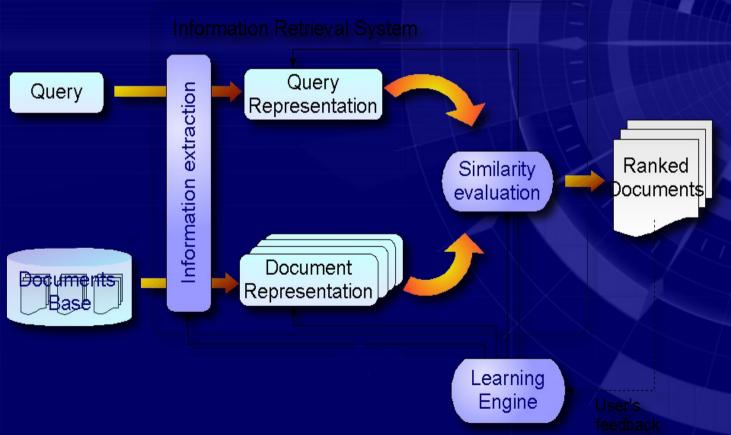


#### **Enhanced IRS with feedback learning**



#### Feedback learning

By giving feedback about the presented document users tell more about their needs to the system





The IRS can enhance itself at multiple levels:

- informationrepresentation
- similarity evaluation

#### **Feedback learning strategies**

#### Many possibilities explored:

- Query rewriting (short term)
- Search context modelling (mid term)
- User model learning (long term)

Proven efficiency of explicit relevance feedback learning concept :

ex: "Rocchio" Algorithm established in the 70's



#### **Explicit vs Implicit feedback**

Experimental (and operational) studies have shown that users are reliable provide explicit feedback on documents

Use of implicit (behavioral) indicators to fill the gap

ex : reading time, scroll behavior, click trough data, ...

#### Implicit data are known:

- to raise privacy issues (but solutions exist)
- to be noisy (or biased, but issues are raised on explicit bias too)
- to be easy to gather in large amount



#### **Explicit vs Implicit feedback**

Hybrid approches combines explicit and implicit data.





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#### Feedback learning and search in context in VITALAS



#### **Analysis of search logs**



Search in context and relevance feedback: starting with search study in collaboration with CWI.

#### Research issues are:

- Do users of professional images IRS have the same behavior as classic users from state-of-the-art studies?
- Are they advanced searchers?
- Can we detect specific behavior pattern?

Aim is to select the right approach.

#### First experiments

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Experiments (conducted by CWI) with "implicit collaboration" us search sessions:

- query suggestion
- term suggestion
- results suggestion

Experiments (conducted by EADS) on using implicit feedback data to infer document interests/relevance:

- interaction events tracking in web based GUI
- framework to learn search context in WebLab platform
- optimization framework for query/term suggestion

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#### Focus on learning using behavior measurements as feed



#### **Search context with feedback**

Explicit and implicit feedback have advantages and drawbacks. It combine feedback through a common framework.

Measurements of current user behavior to extract interests:

- Time spent on reading a document
- Selection of terms in abstract
- Click on a link after reading its description
- Explicit rating of items

Matrix X of measurements per documents/parts of documents

X(i,j) = measure of behavior j on element i

$$X = \begin{vmatrix} m_{1,1} & m_{1,2} & m_{1,3} & m_{1,4} \\ m_{2,1} & m_{2,2} & m_{2,3} & m_{2,4} \\ m_{3,1} & m_{3,2} & m_{3,4} & m_{3,4} \end{vmatrix}$$



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#### Search context with implicit feedback

Using past search history to learn relevance pattern in behavior measurements

Rel= matrix of behavior patterns on relevant documents

$$Rel = egin{bmatrix} r_{1,1} & r_{1,2} & r_{1,3} & r_{1,4} \ r_{2,1} & r_{2,2} & r_{2,3} & r_{2,4} \ r_{3,1} & r_{3,2} & r_{3,4} & r_{3,4} \ \end{pmatrix}$$

Irr = matrix of behavior patterns on irrelevant documents

 Classic supervised learning problem which enables the computation of current search context.

ex: a weighted vector of terms reflecting current interests.

#### **Searching in context**

Using the search context to enhance user experience while in a session:

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- Query expansion and/or suggestion to help users to define their needs
- Changing the similarity and ranking algorithm to personalize the behavior of the system to the user and its current needs
- Adapting the presentation of results
- Providing tools to interact/explore the corpus (to provide more accurate data for implicit relevance feedback)

#### Searching in context: a multi objective optimisation probl



Query expansion and/or suggestion as a multi objective optimization problem: : finding the "best query" regarding multiple criteria and constraints

$$\vec{f}(\vec{x}) = \left\{ f_1(\vec{x}), \dots, f_i(\vec{x}), \dots, f_n(\vec{x}) \right\}$$
with  $\vec{g}(\vec{x}) \ge 0 \Leftrightarrow \left\{ g_1(\vec{x}) \ge 0, \dots, g_i(\vec{x}) \ge 0, \dots, g_m(\vec{x} \ge 0) \right\}$ 

Criteria examples: Precision, Recall, Diversity, Novelty

$$P = rac{N_{relevant\,results}}{N_{results}} \hspace{1cm} R = rac{N_{relevant\,results}}{N_{relevant\,doc}}$$

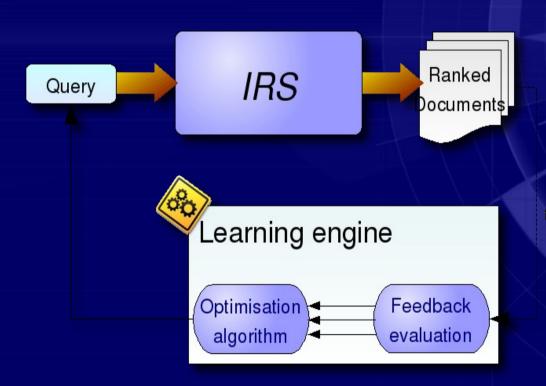
Adapted and personalized for each user or community of users

#### **Evolutionary algorithm for query expansion/suggestion**

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Evolutionary algorithms to optimize the first user query:



- 1. Given a query of N terms
- 2. Rank document
- 3. Evaluate criteria through user's feedback
- 4. Optimise the query vector to maximize the criteria

#### **Evolutionary algorithm for query expansion/suggestion**

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#### Difficulties:

- High dimensionality of term space
   reduced through the use of search context learned from feedback
- Combinaison advanced query operator
   use of genetic programming to compute advanced queries

#### Multiple level of impact:

- Query suggestion (with or without complex syntax)
- Term suggestion to disambiguate with context
- Implicit rewriting with "push" of new documents

#### **Evolutionary algorithm for query expansion/suggestion**



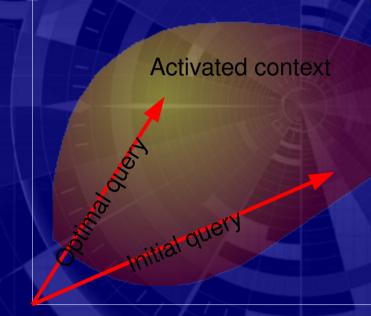


User query vector in the whole vocabulary of N terms

$$ec{Q} = egin{array}{c} word_1 - 0 \ word_2 - 1 \ word_3 - 0 \ ... \ word_N - 1 \ \end{array}$$

Search context which "activate" some parts of the vocabulary

$$ec{C} = egin{array}{c} word_1 - 0,0 \ word_2 - 0,89 \ word_3 - 0,5 \ ... \ word_N - 0,01 \ \end{array}$$



Geometrical representation of search space, initial user query and search context used to limit the searched area

#### **Expanding context using semantic and collaboration**



Use of semantic knowledge bases to expand the context : changing information model to concept space

- Classification/clustering problem in a graph or a hierarchy of semantic concepts
- Use of Word Sense Disambiguation (WSD) techniques

Knowledge representation comes out users past searches: use of collaborative search experiences and/or external bases (ontologies, wikipedia...)



#### **Conclusion and future work**



#### **Future research paths**

- Finalisation of Logs analysis
- Exploration of feedback learning approaches
  - Implicit relevance learning with already existent data: Experimentation of state-of-the-art approaches in IR based on statistics and of pattern recognition approaches
  - Query expansion with evolutionary algorithms to optimise query weights with operators
  - Extension of collaboration to enlarge user search context by using user model similarity matching and developing new paradigm of collaboration
- Evaluation of proposed approaches
- Integration within WebLab platform (VITALAS V2 ?)



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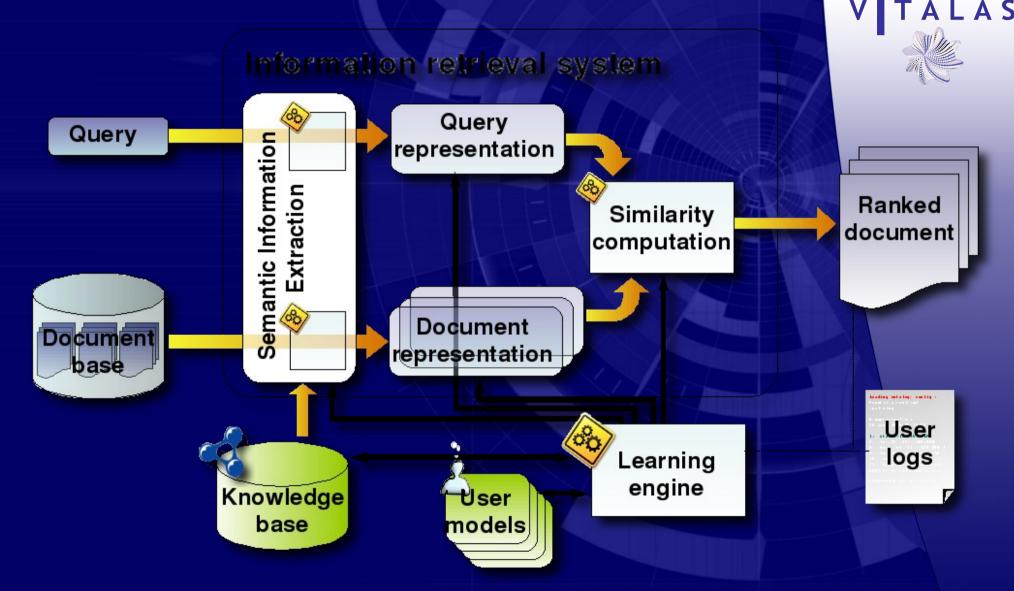
#### About the author

- Research engineer at EADS DS
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- Involved in VITALAS (EC project 2007/2009)
  - EADS DS as software architect
  - Personal involvement in "search in context"
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#### **Enhanced IRS**



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