Semantic Web Technologies for Capturing, Sharing and Reusing Knowledge

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Sharing

The Knowledge Life-cycle

Why manage knowledge?

- 5
- To enable easy <u>timely and effective</u> reuse
 - We need: to enable sharing
 - Requirements: easy and effective sharing
- To enable sharing
 - we need to: capture knowledge
 - Desiderata:
 - Easy capture (do not get in the way of the user's work!)
 - Comprehensive capture (do not miss important facts!)
- To enable capture:
 - We need acquiring and modelling the domain and process it in an appropriate way

Please note: most books and tutorial work the other way around. They start with modelling (e.g. ontology building) then move to acquisition, then to sharing (if they do!). This often generates confusion: modelling seems the most important issue!!

Today's tutorial



- We will see techniques and methodologies for
 - Knowledge Capture
 - Extracting and integrating information
 - from existing archives and documents
 - With user in the loop
 - Knowledge Sharing and Reuse
 - Enabling knowledge searching
- You have already seen:
 - Knowledge Acquisition and Modelling
 - Ontology Engineering

Ontology rdfs:domair rdfs:range Person dfs:subClassOf rdfs:subClassOf Academic Staff Graduate rdfs:subClassOf rdfs:subClassO PhDStudent Lecturer rdf:type rdf:type wre: PhDStudent swic Lecturer if:about="http://www.aifb.uni-karls rdf:about="http://www.aifb.uni -karlsruhe.de/WBS/sst/ ruhe.de/WBS/sha/#Siegfried"> swrc:nmme>Siegfried Handschuh #Steffen" 000 swrc:name> Anno-IPAS Search Engine | University of swrc:cooperatesWith rdf;resource . tation swrc:cooperatesWith < 5513 "http://www.aifb.uni-karlsruhe.de /WBS/sst/#Steffen"/> /swrc: swrc:PhDStudent Steffen Siegfried Web Handschub 2 Page Research erantic We URL ttp://www.aifb.uni-karlsruhe.de/MBS/sha Inding [1] Event Report Data Trent type: has_Engine_Type gapore) FSO Sear EEC suite case loom 1 channel A EEC suite case loom 1 channel B Done cancel <<Less id not get to sample the reddish h sdsds sdsds michaling on EEC Comment: tate&url=file: /home/inas/inas/html/AnonEventP this version of the loom is also used on AK-234-I omment 化外化 Semantic Web for **Knowledge** Capture

cooperatesWith>

Knowledge Capture

- Collecting and aggregating knowledge within and across media / archives
 - in a rich, semantically-oriented way
- Two main tasks
 - Annotating existing structured resources
 - E.g. Databases or dictionaries
 - See Guus' talk
 - Annotating unstructured documents
 - Texts, images, data, etc.
 - This tutorial

Semantic Web for KC from Docs

- Two moments to capture knowledge
 - At source: helping people capturing knowledge when produced
 - On legacy documents, pictures, data: - Annotation services
- Outcome of capture
 - A semantic representation of (part of) the content
 - Enrichment of multimedia documents
 - with layers of manually or automatically generated annotation

Compound Documents & kC

- Typical data objects (text, image, raw data)
 - Text formats: Word, Excel, PPT and PDF documents
 - Images: Jpeg and Gif
 - Raw data: Measurements stored in a RDBMS
 - Cross-media: Compound documents: Word, PPTs and PDFs containing both text and Jpeg images
 - Portions semantically related to each other within the same physical document
 - Information contained in just one modality is insufficient
 - Cross-media knowledge acquisition techniques needed in order to capture and manage all of the explicit and implicit knowledge





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Stripping words off documents takes you nowhere

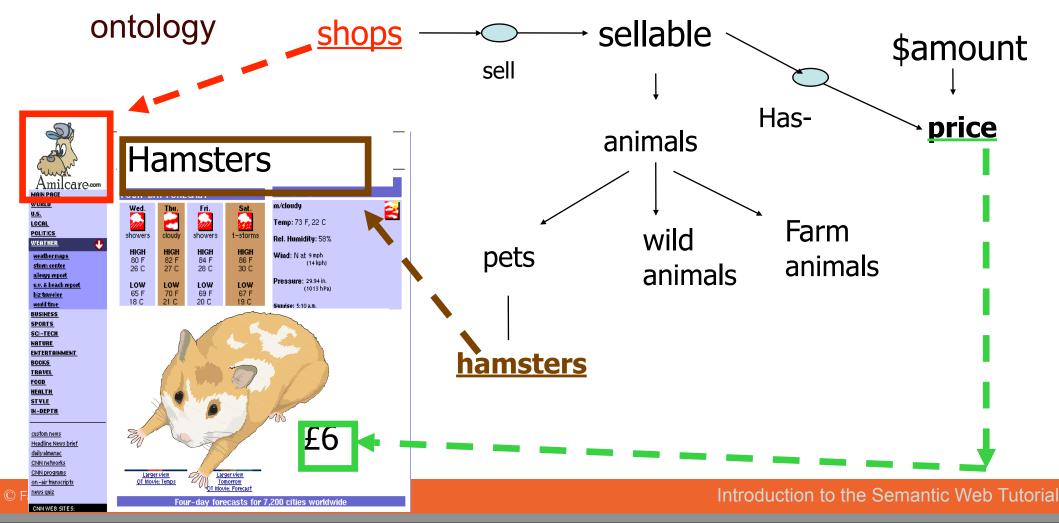
Ontology-based annotations

- Allows:
 - Ontology-driven processing
 - Services based on ontology will be able to use information

Ontology-based Annotation

Marking up contained information

Portions of documents associated to objects in

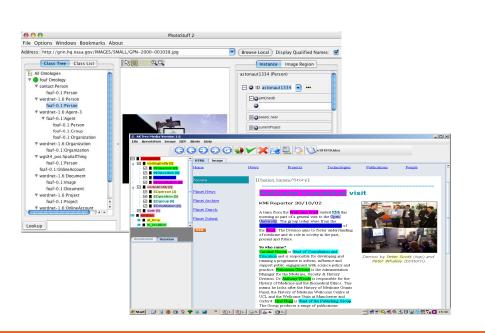


Input & Output

- Input to the KC technologies
 - Ontologies (MMO, domain ontology),
 - Background knowledge (gazetteers, etc.)
 - Documents
 - Possiby in normalised form
 - Medium to extract from (text, images, data, videos,...)
- Output
 - Annotations (e.g. RDF triples stored in triple store)
 - May be in the form of uncertain output

Some Useful Tools

- 5
- User-friendly tools for annotation
 - Cream (Handschuh et al. 2002)
 - Melita (Ciravegna et al. 2002)
 - Photostuff (Hendler et al. 2005)
 - AktiveMedia (Chakravarthy et al. 2006)



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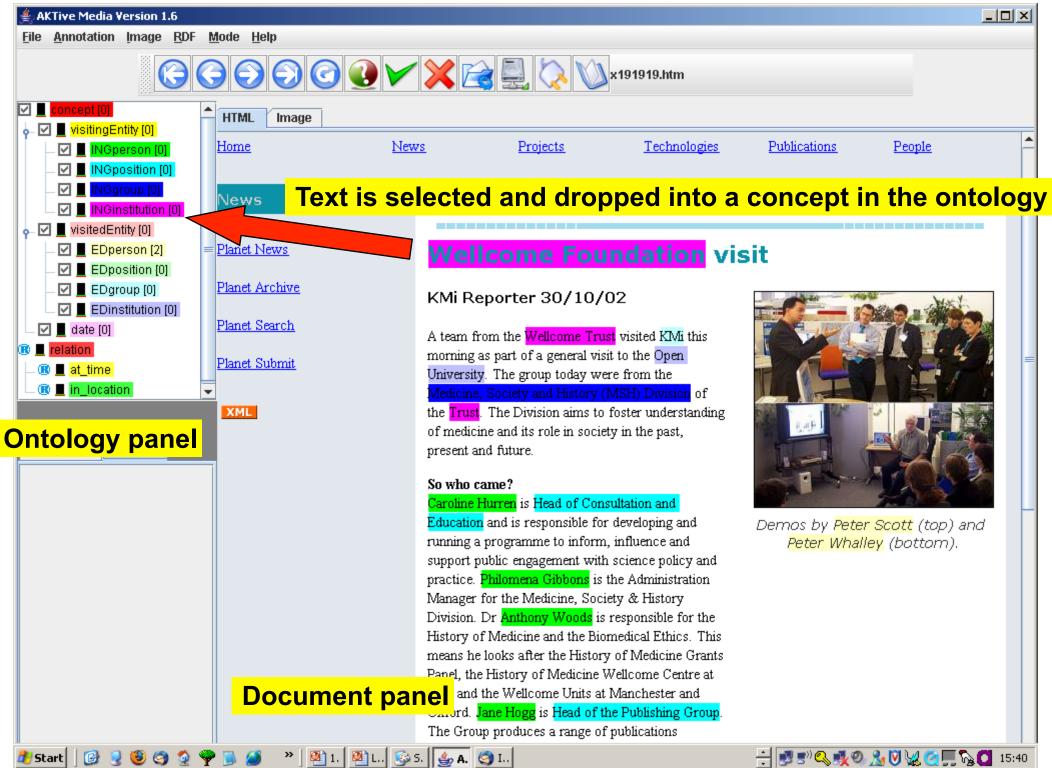


AktiveMedia



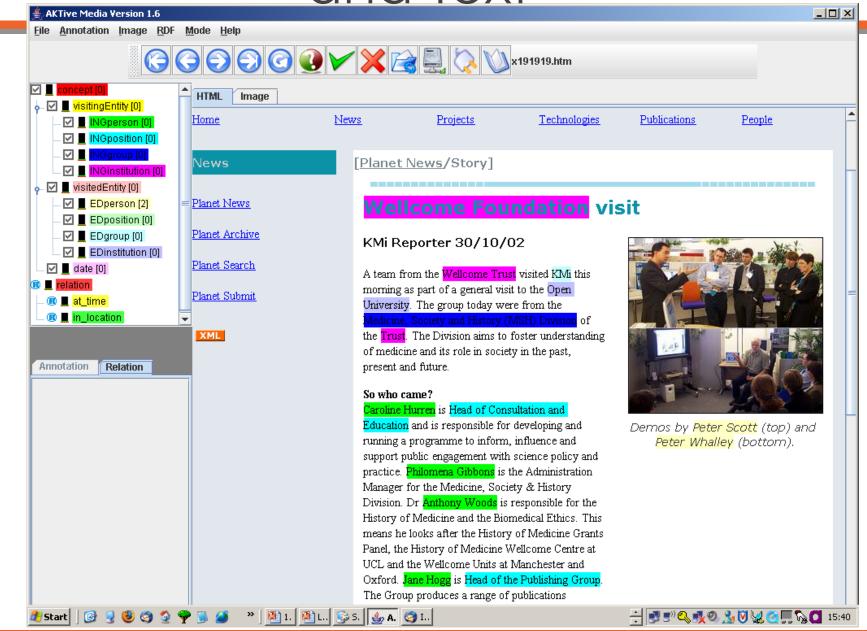
- Enables semi-automatic annotation across texts and images
- The interface enables
 - Annotation of documents in RDF based on an OWL ontology
- Types of annotations
 - Concepts / Relations
- SW: Annotation:
 - Selection of concept/relation and highlighting of text is the way in which annotation is performed

http://www.dcs.shef.ac.uk/~ajay/html/cresearch.html



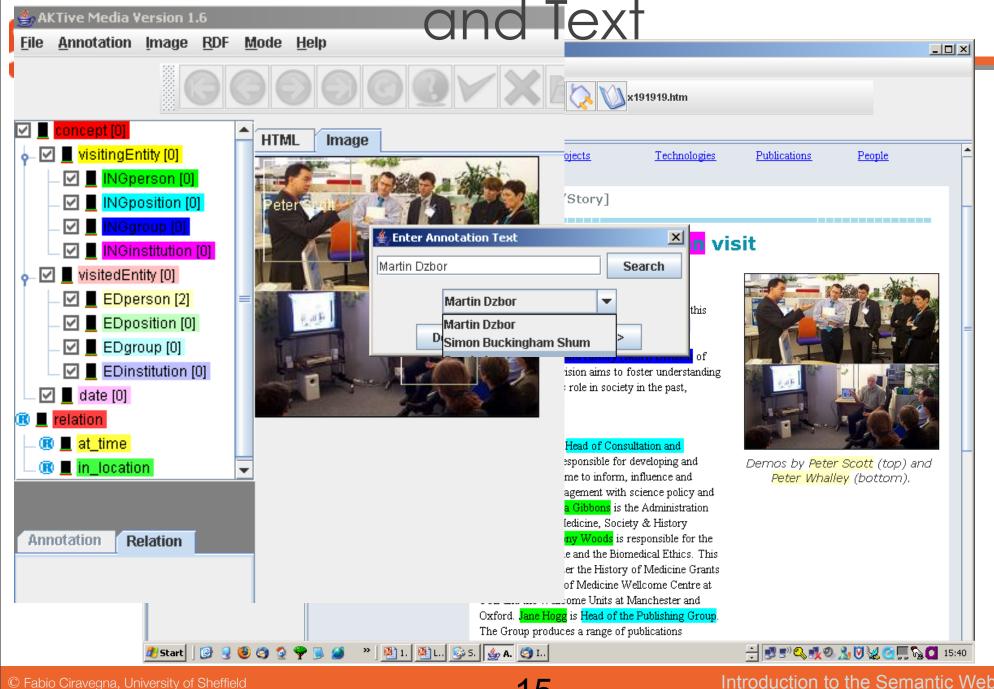
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Contextual Annotation of Images and Text





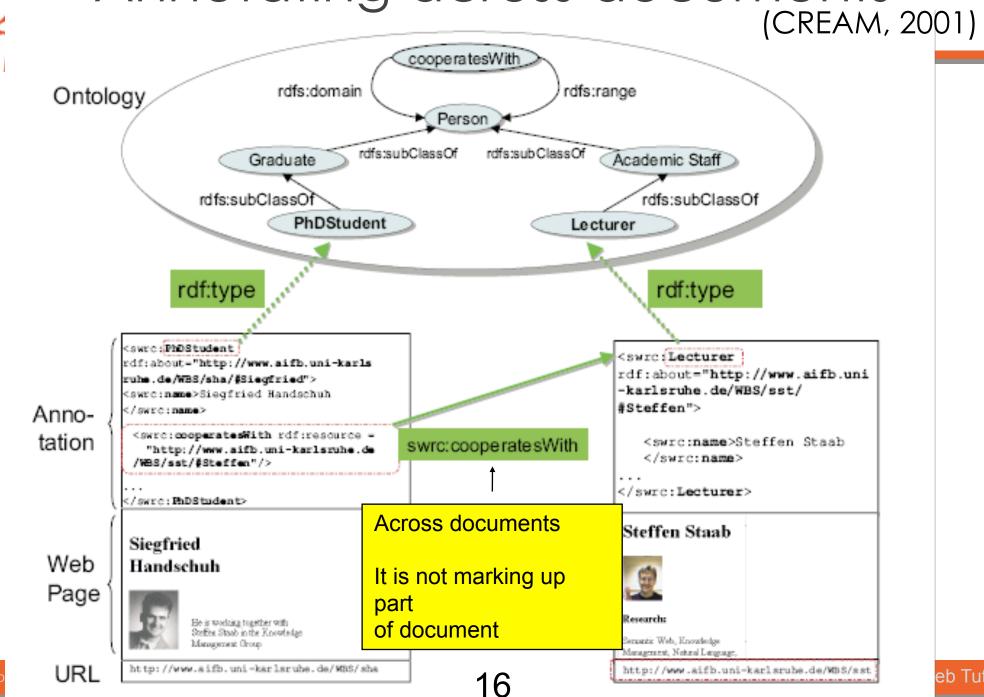
Contextual Annotation of Images



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Introduction to the Semantic Web Tutorial

Annotating across documents



Marking up Provenance

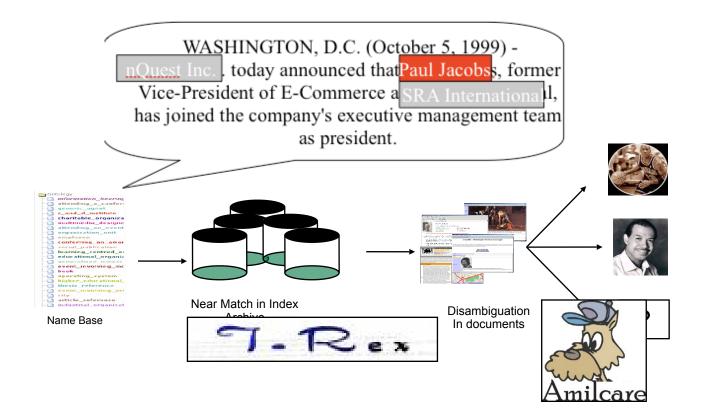
- COMM A Core Ontology for Multimedia based on
 - the MPEG-7 standard

http://comm.semanticweb.org/

- the DOLCE foundational ontology.

```
<Mpeg7>
```

```
<Description xsi:type="ContentEntityType">
<MultimediaContent xsi:type="ImageType">
  <Image id="IMG1">
   <SpatialDecomposition>
    <StillRegion id="SR1">
    <Semantic>
      <Label><Name> Roosevelt </Name></Label>
    </Semantic>
    </StillRegion>
    <StillRegion id="SR2">
     <TextAnnotation>
                          <!-- TextAnnotationType -->
      <KeywordAnnotation><Keyword> Churchill </Keyword></KeywordAnnotation>
    </TextAnnotation>
    </StillRegion>
    <StillRegion id="SR3">
    <Semantic>
      <Definition> <!-- Also TextAnnotationType -->
      <StructuredAnnotation><Who><Name> Stalin </Name></Who></StructuredAnnotation>
      </Definition>
     </Semantic>
    </StillRegion>
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Automating Annotation

Annotation Engines

- 5
- Solutions like AktiveMedia can be used for annotating new documents and knowledge
 - large repositories of legacy data exist
 - it is important that new management solutions are able to reuse existing data
 - Be humble: do not require a completely new world to be built for you!!
- Legacy data is generally represented in
 - databases
 - textual documents
 - images

- ...

Tasks for KA: Extraction

- Text:
 - Entity Extraction
 - Table Fields Extraction
 - Relation Extraction
 - Event Extraction
- Data:
 - Similarity of Data Instances
 - Functions and relation
 - Finding patterns and (ir-)regularities in data

- Images:
 - Semantically driven Image analysis using ontologies, for retrieval and annotation
 - Image classification/ clustering with respect to the dominant visual trends

- Automatically extracting pre-specified information from textual documents
 - salient facts about pre-specified types of events, entities or relationships.
- Populating a structured information source from a semi-structured, unstructured, or free text, information source.





Growing complexity



- salient facts about pre-specified types of events, entities or relationships.
- Populating a structured information source from a semi-structured, unstructured, or free text, information source.

WASHINGTON, D.C. (October 5, 1999) nQuest Inc. today announced that Paul Jacobs, former Vice-President of E-Commerce at SRA International, has joined the company's executive management team as president.

Named Entities

Event Recognition

Growing complexity

- Automatically extracting pre-specified information from textual documents
 - salient facts about pre-specified types of events, entities or relationships.
- Populating a structured information source from a semi-structured, unstructured, or free text, information source.





Growing complexity

- Automatically extracting pre-specified information from textual documents
 - salient facts about pre-specified types of events, entities or relationships.
- Populating a structured information semi-structured, unstructured Company: nQuest Inc. Date: today source.

InPerson: Paul Jacobs InRole: president

Company: SRA International **OutPerson**: Paul Jacobs **OutRole**: Vice-President of E-Commerce,

Named Entities

Growing complexity

Event Recognition

Classic Tasks



- Entity Extraction
- Fields Extraction
- Relation Extraction
- Event Extraction
- Other (non Semantic) Tasks
 - Document Similarity
 - Text Categorization

Named Entity Recognition

- Tasks:
 - Recognition and classification of named entities
 - E.g. people's names, companies, locations, etc.
 - Unique identification of named entities (URI assignment)
 - Including disambiguation
 - Michael Jordan as basketball player Vs lawyer
 - London UK Vs London USA
 - Integration with other sources
 - E.g. positioning on a map

Traditional approach to NER&C

- Two steps:
 - Training phase
 - Input: annotated set of representative documents
 - Output: trained system
 - At runtime
 - One-by-one document analysis
- Expected accuracy:
 - 80-95% (free texts)
 - Web documents tend to require additional processing to get equivalent results (but doable to some extent)
- Medium Scale: up to hundreds of thousands of documents

Large Scale NER&C

- For large scale (some hundred millions pages) smarter infrastructure is needed
 - Search engine-like indexing infrastructure
 - Faster processing (less processing)
 - -Two cases:
 - Recognition of known terms (and their variations)
 - See also information integration
 - Discovery of new names

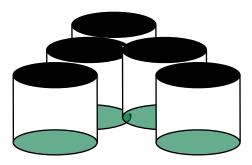
S. Dill, N. Eiron, et al: SemTag and Seeker: Bootstrapping the semantic web via automated semantic annotation. WWW'03

Large Scale NER: Indexing



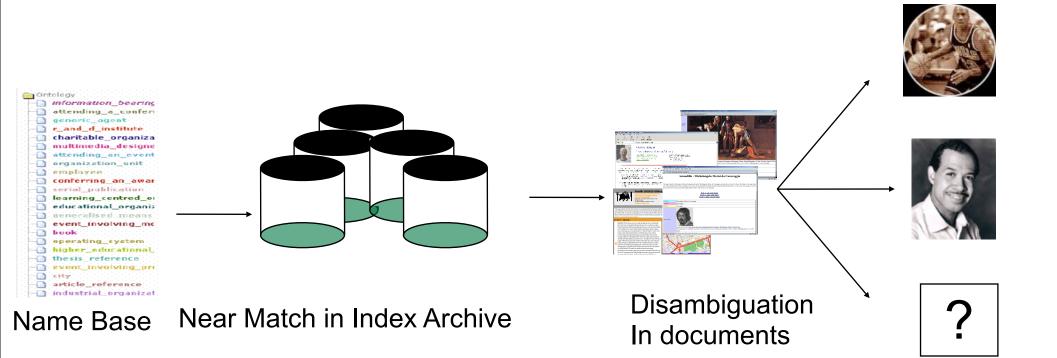
Document Indexing as in Search Engines





Distributed Index Archive (keywords)

Known Name Recognition



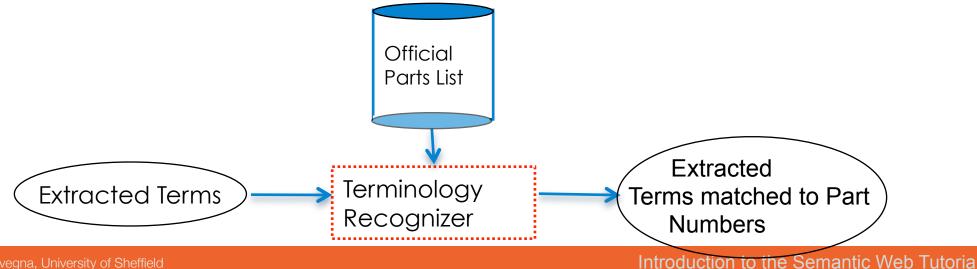
S. Dill, N. Eiron, et al: SemTag and Seeker: Bootstrapping the semantic web via automated semantic annotation. WWW'03

Discovery of New Names

- Modified Indexing of documents to recognize potential names
 - Traditional NER
 - On the window of words (not the whole doc!!!)
 - Fast and effective
 - Web specific strategies
 - To identify names without context

Terminology Recognition

- NER is one example of term recognition
- More useful in technical domains is terminology recognition
 - The task of assigning a URI to a technical description
 - i.e. mapping a natural language description to the official company ontology



Example of TR

- LP (FAN) COMPRESSOR BLADE
- LP COMPRESSOR SPLITTER FAIRING
- LOW PRESSURE COMPRESSOR BEARING
- HIGH PRESSURE COMPRESSOR BEARING
- SPLITTER FAIRING LP/IP COMPRESSOR
- 1-6 COMPRESSOR ASSEMBLY
- AIR DUCT OUTLET IP COMPRESSOR
- BLADE STAGE 6 HP COMPRESSOR

Query = "Low Pressure Compressor Fairing"

Table Field Extraction

- Tables are an essential part of many documents
 - Most information is represented in tables
- Tables can be represented as forms to fill
 - Semantics is fixed
 - Wrapper writing or wrapper induction (Kushmerick 1997)
- Tables can be created ad hoc in documents (e.g. Word docs)
 - Semantics is unclear
 - Sometimes documents are created as part of a workflow, therefore they tend to be created using common models
 - e.g. by re-using the previously generated document
 - hence tables evolve, but still semantics can be traced

More complex IE: event modelling

- 5
- Not just NER but also relation among elements in a document
 - More complex task
 - Requires some reasoning to bridge the complexity of events to the ontology structure
 - Imprecision in extraction
 - Information non matching the ontology schema
- This is where IE has hit a performance ceiling
 - 60/70 Precision/Recall ratio since 1998

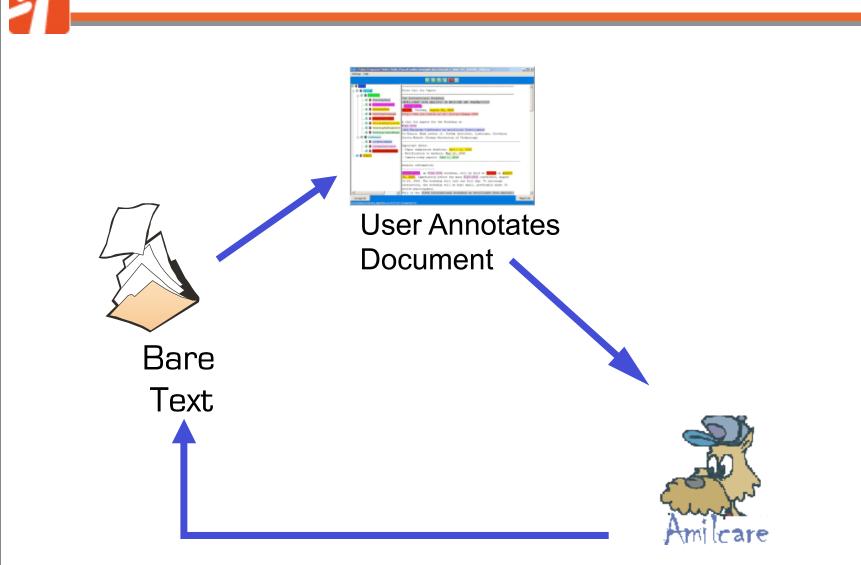
A list of tools for automatic annotation

- Architectures for IE:
 - UIMA (http://www.research.ibm.com/UIMA/)
 - GATE (<u>www.gate.ac.uk</u>)
 - Contains Annie: Named Entity Recogniser
 - KIM (http://www.ontotext.com/kim/)
- WiT toolbox: http://nlp.shef.ac.uk/wig/tools/)
 - Manual and semi-automatic annotation of texts and images
 - AktiveMedia http://www.dcs.shef.ac.uk/~ajay/html/cresearch.html
 - TRex: plugin for Machine Learning based IE http://tyne.shef.ac.uk/t-rex/index.html
 - Saxon: rule-based (FST) tool

http://nlp.shef.ac.uk/wig/tools/saxon/



Using IE to support annotation: step 1

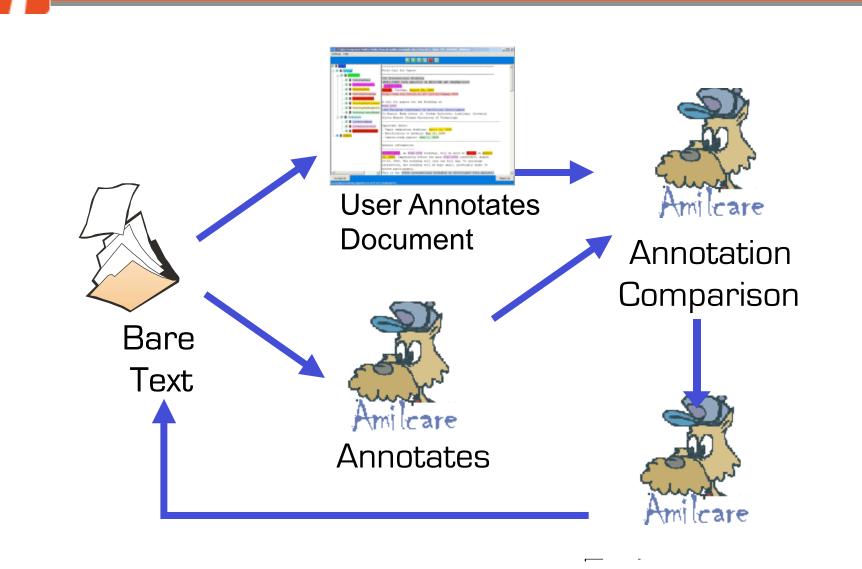


Trains on annotated corpus



Using IE to support annotation: step 1

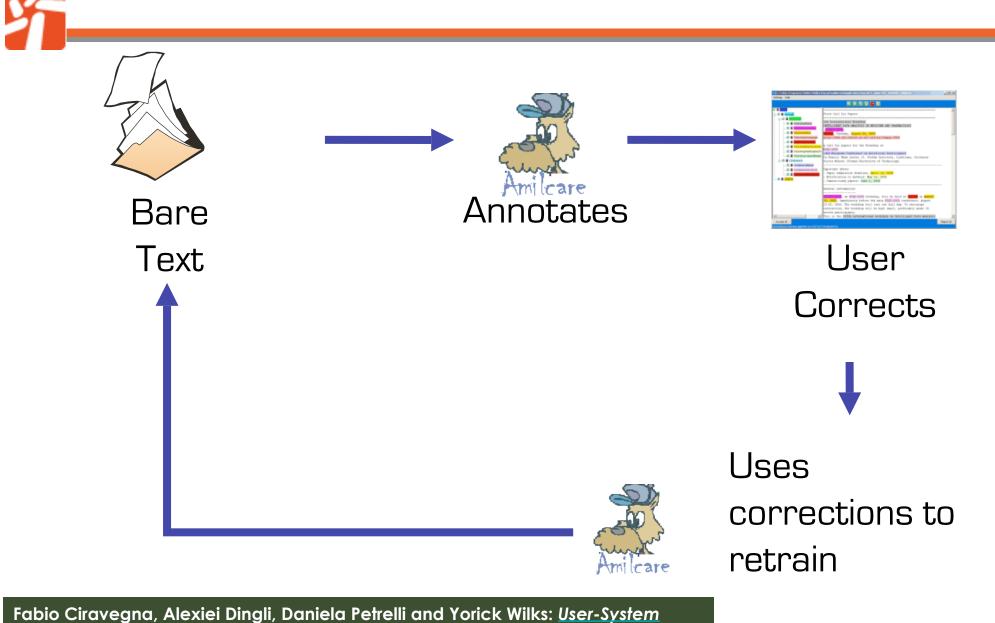
34



Retrain using errors,

missing tags and mistakes

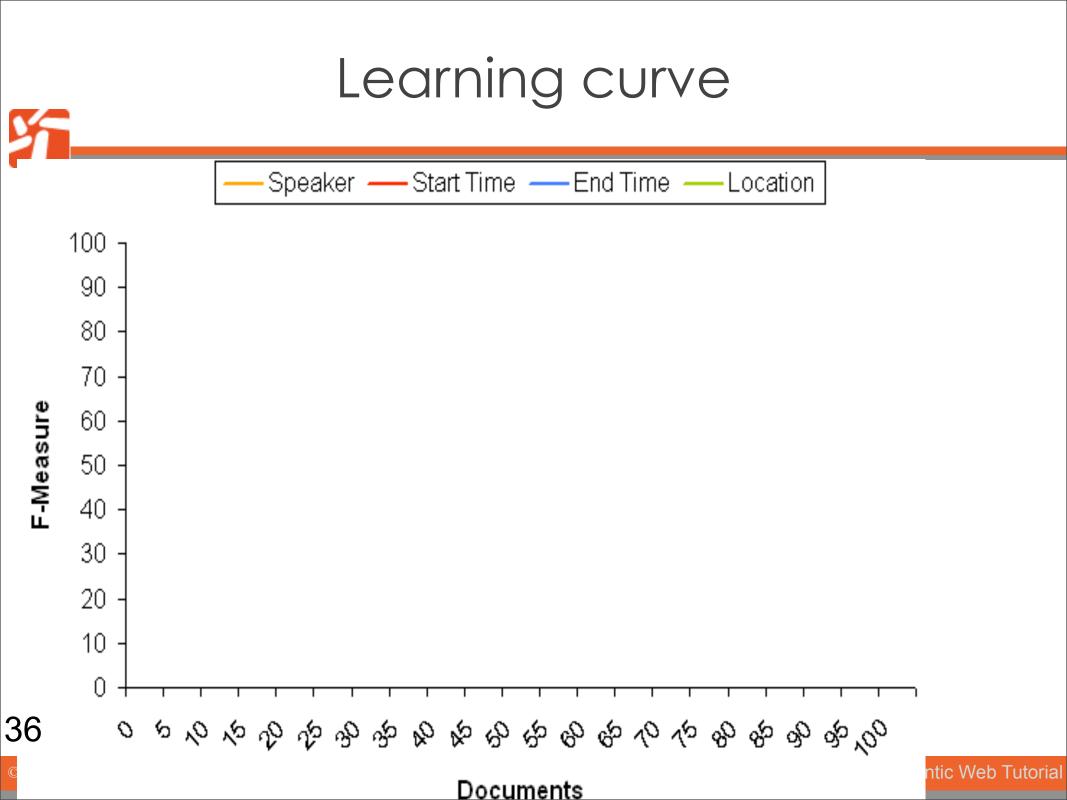
Using IE to support annotation: step 2

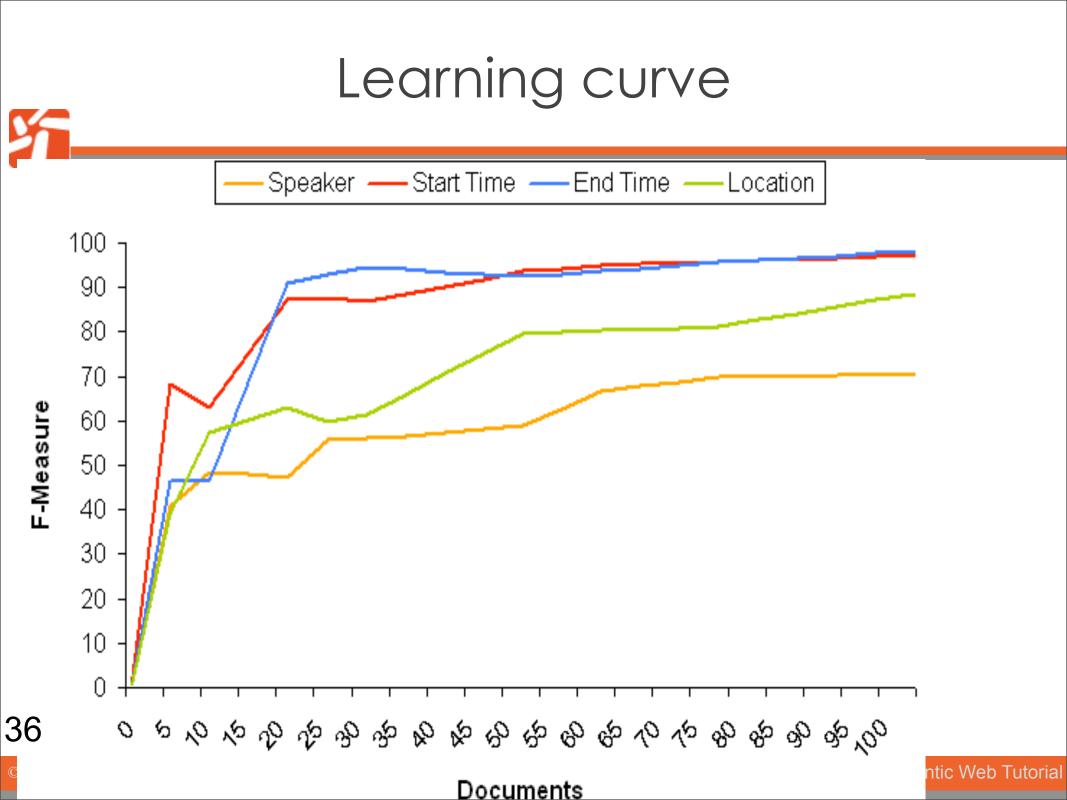


<u>Cooperation in Document Annotation based on Information Extraction</u>, EKAW 2002

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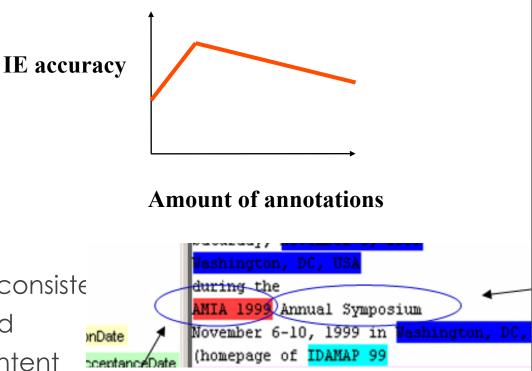
Introduction to the Semantic Web Tutorial

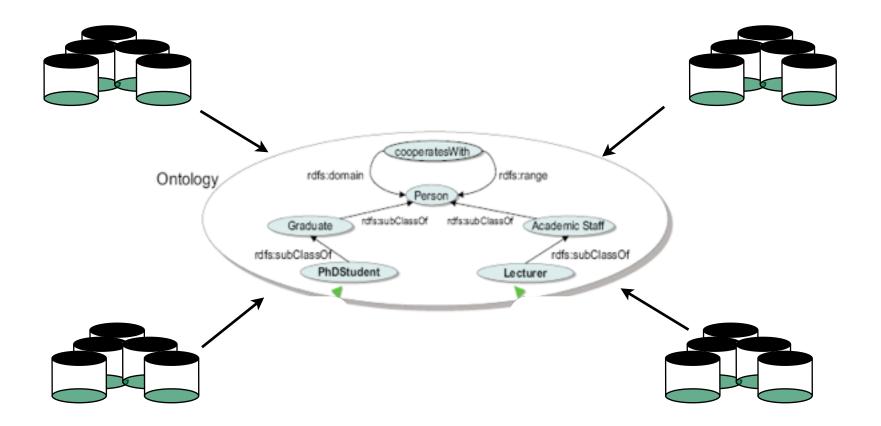




Impact on Annotation

- University of Karlsruhe experiments
 - -80% annotation time
 - +100 interannotator agreement
 - Is this positive?
- Outstanding issue:
 - Impact on annotators of suggestions topping 85% accuracy?
 - Annotation needs to be precise and consiste
 - Otherwise the IE system is confused
 - Can only annotate document content
 - With connections to the rest of the knowledge via information integration





Information Integration

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Information Integration

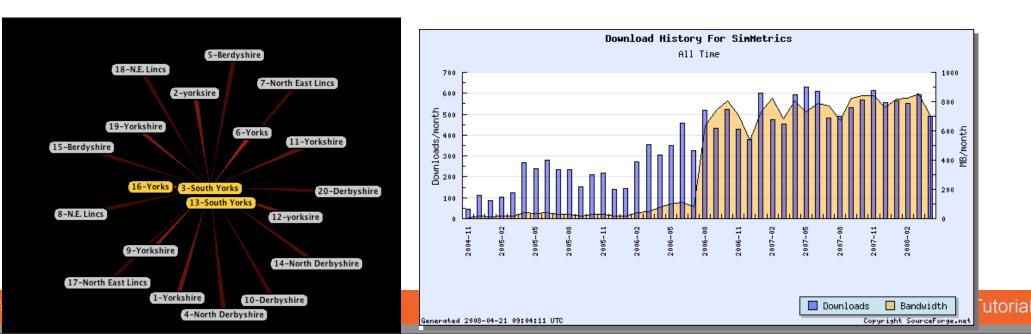
- Facts from different sources need to be integrated
 - To connect information/knowledge across docs
 - Assign unique URI
 - To solve discrepancies and ambiguities
- Steps
 - Unique instance identification (for entities)
 - Record linkage (for events)
- Information Integration strategies
 - Generic
 - Distance metrics (Chapman 2004)
 - Using Web bias

- Statistical matching
- Application specific
 - Rules

SimMetrics

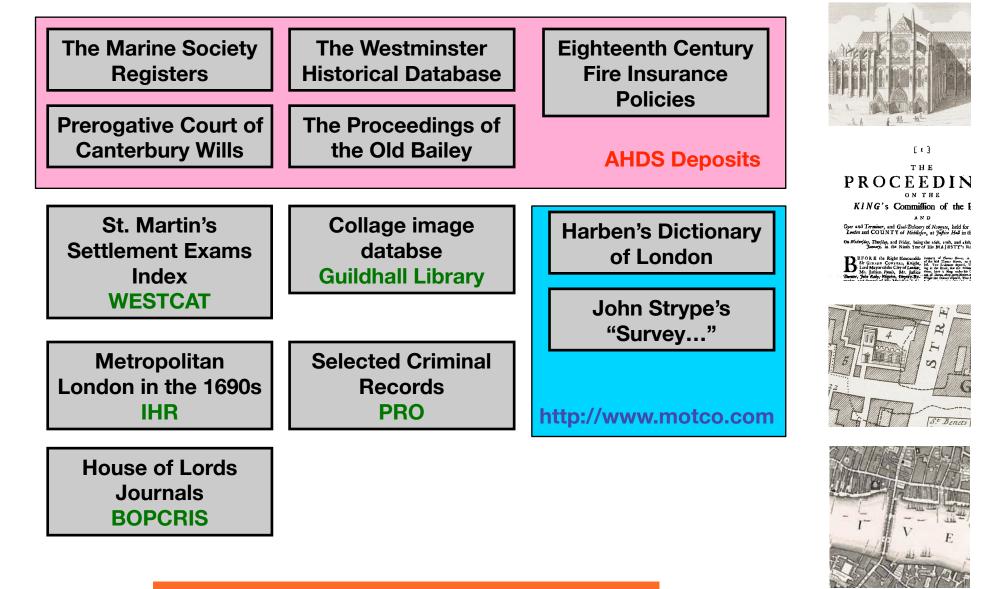


- Library of distance metrics released as open source
 - <u>http://sourceforge.net/projects/simmetrics/</u>
 - >15,000 downloads since end of 2004
 - Most downloaded distance metrics library on the Web
 - for strings and records
 - Hundreds of applications
 - Developed by Sam Chapman, University of Sheffield





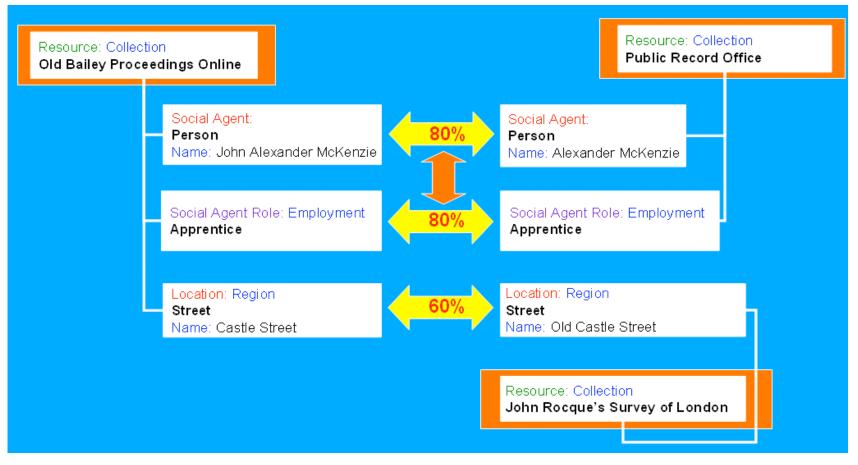
Armadillo: Historical Data Mining



http://www.hrionline.ac.uk/armadillo/

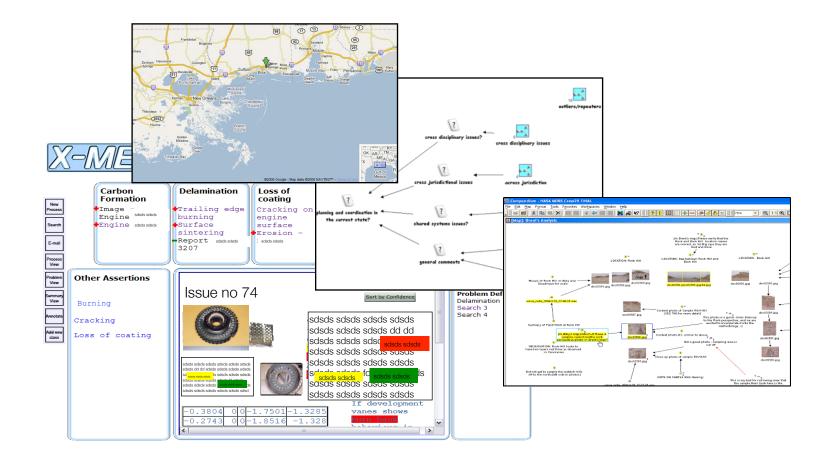
Information Integration

Armadillo: Historical Data Mining





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Sem Web for Knowledge Sharing and Reuse

Knowledge Sharing and Reuse

- In KM mainly means
 - Retrieving information and knowledge
 - At the right time
 - In the right form
 - » E.g. independently from where it is stored
 - » Or even the form in which it is stored
 - Suitable to the specific users
 - » e.g. patients should net receive information using technical terms
 - Suitable to specific interests
 - » I am working on social aspects of SW, not interested in engineering aspect of SW
 - In an efficient and effective way
 - Coping with large scale
 - Supporting processes

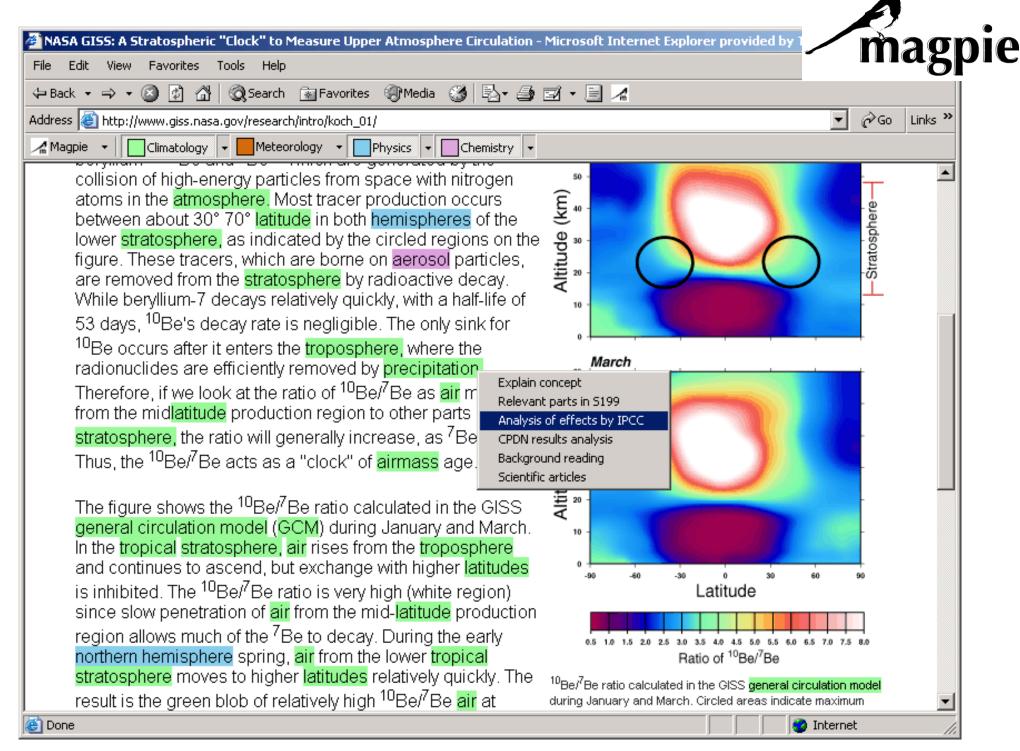
Sharing and Reuse via SW

- Ontology based annotation enables
 - Searching using ontologies
 - Searching metadata rather than text
 - Connection of information across documents, media and archives
 - Retrieving information independently from the store/ media
 - Reasoning on knowledge
 - Making implicit explicit
 - Workflow support
 - Supporting user actions rather than single searches

Document enrichment

- Adding knowledge to documents (ctd.)
 - Document enrichment: helping connecting the document to the rest of the knowledge
 - Associating Services
 - Magpie (Dzbor et al. 2004)
 - Connected to other documents

 e.g. Automatic generation of hyperlinks
 COHSE (Goble et al. 2001)



Searching using Sem Web

- Many types of technologies
 - Search based on structural query languages, such as SPARQL, see, e.g., ARQ, and
 - User-centred search to retrieve
 ontologies (e.g. Swoogle [Ding et al.
 2004] and Watson [d'Aquin et al. 2007])
 - User-centred approaches to retrieve information and knowledge
 - We will see the latter

Why Search?

- Task in Searching
 - Document Searching (images/texts/videos/ data)
 - Goal of query is to retrieve documents
 - Semantic Search is used as replacement for traditional keyword based systems
 - Knowledge Searching
 - Goal is to retrieve knowledge (i.e. triples)
 - This is similar to search a virtual database
 - Independently from source
 - » Documents can be accessed for
 - » Provenance analysis (checking correctness)
 - » Further browsing

Keywords Based Search

- 5
- Document search:
 - Two main issues,
 - Ambiguity:
 - Keywords can be polysemous, i.e. they can have multiple meanings.
 - » Search returns spurious documents (low precision)
 - Synonymity:
 - an object can be identified by multiple equivalent terms
 - » Search does not return documents containing other synonyms (low recall)
- Knowledge search
 - Not supported

Semantic Search (OS)

- Search instances in an ontology
 - Annotations are unambiguous
 - OS Does not suffer from ambiguity and synonym issues of keyword-based systems (KS)
 - Supports knowledge search
 - Naturally
 - Supports document search
 But...

User Centred Approaches

- By merging the definitions in [Uren et al. 2008], [Kaufmann et al. 2007b] and [Baghdev et al. 2008]:
 - Keyword-based approaches considering a natural language query as a bag of words
 - [Kaufmann et al. 2007a] [Lei et al., 2006])
 - Natural language approaches: modelling the linguistics of the query
 - [Lopez et al. 2005],[Bernstein et al. 2005b], [Kaufmann et al. 2006]
 - Graph-based approaches
 - [Bernstein et al. 2005a], SEWASIE, Falcon-S.
 - Form-based approaches (e.g. Corese)
 - Hybrid approaches
 - K-Search [Baghdev et al. 2008])

Semantic Search Approaches (1)

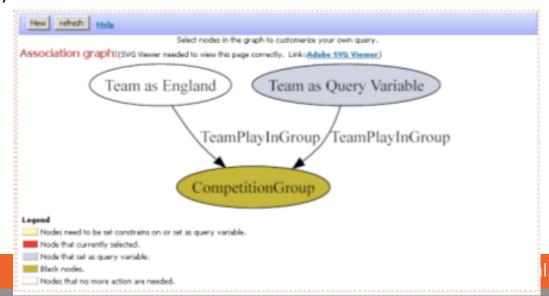
- Keyword-based approaches
 - Query via keywords
 - <u>All</u> the keywords are mapped to Semantic Concepts
 - Requirements: feedback on generated query
 - Issues:
 - User lost for words
 - What is covered by the ontology?
 - What if my keyword does not map to ontology?
- E.g. SemSearch

lome <u>Knowledge Sources</u> Tools Ontologies					
nantic Search					
nis search engine searches relev draction tool ASDI. User can ad	vant data from the back-end semantic data repository extracted by our meta-data d a subject to narrow down queries by using format like "subject:keyword".				

how search summary Refine search

Semantic Search Approaches (2)

- View-based approaches
 - Based on querying by building visual graphs
 - Advantages:
 - What covered by ontology is always clear
 - Search is intuitive and liked by users
 - Issues
 - Can be fairly rigid and constraining
 - Kaufmann et al 2007 report a very high time required for querying
- E.g. Falcon



Semantic Search Approaches (3)

- A natural language approach
 - Interprets fully fledged NL questions
 - Requirements:
 - Feedback on generated query
 - Issues:
 - User lost for words
 - What is covered by the ontology?
 - NL can be tricky (limited linguistic coverage)
- E.g. Aqua

Ask a Show	me all planet stories writter arning Mechanism f		Askl	Exam	ples	LOGIN	ou are logged a: anonymou:
Relation Similarity Se Query Validated Logical Representat	Category WH_3TERM	lation - Second Term - Third Terr	n.				
Note: The Lexicor	kmi-planet-news-item (learning mechanism) researcher	written <u>has-author owned-by</u> mapping to (has-author owned <u>has-project-member has-project</u> is mapping to (has-project-memi	- E I-by) t-leader - a	ikt	- <u>akt</u> -	[WH_STERM] [WH_UNKNREL]	

Semantic Search Approaches (4)

- Form-based approaches
 - The ontology is turned into a form and queries are expressed by filling conditions into the form
 - Advantages:
 - What covered by ontology is always clear
 - Issues
 - Can be fairly rigid and constraining

Search	Corporate Knowledge
Gol	Connect 🗆 No Join 🗆 Style sheet std 💌 Search More Rule Clear
(Advanced search)	Team
	Properties
De 1000 1	Profession engineer group
Query	Skill java programming
Team	Profession researcher group
Apply	Skill HCI
Table	Profession manager group
Skil	
Skil	Skill none

Ontology-based Querying: Issues

- Metadata may cover only partially the user information needs
 - Limitations in the ontology wrt user needs
 - Often the use people will do of information is impossible to foresee
 - Limitations in the annotation capabilities
 - Sometimes Information is impossible to retrieve reliably using automatic methods
 - Metadata unavailable for a specific document

An Experiment on Jet Engine Event Reports

- 21 topics of search, e.g.
 - "How many events were caused during maintenance in 2003?"
 - "What events were caused during maintenance in 2003 due to control units?"
 - 'Find al I the events associated with damage to acous- tic liners fol lowing bird strike''
- How many topics can we model with Information Extraction?
 - 21 topics/ 14 topics partially or not covered by IE-based annotations
 - given size of corpus there is no way that manual annotations are added

Issues and Solutions

- Ontology can be extended
 - But increases effort in indexing
 - Equivalent to extending metadata in SDM
 - But it is impossible to foresee all uses of information
 - Ontology will always be insufficient somehow
 - Information Extraction can be used to reduce burden of annotation

- But some parts are irretrievable

Hybrid Search

- [Bhagdev et al 2008] propose a model of searching combining
 - the flexibility of keyword-based retrieval
 - querying and reasoning capabilities of semantic search
- HS is formally defined as:
 - the application of semantic (metadata-based) search for the parts of the user queries
 - where metadata is available
 - the application of keyword-based search for the parts not covered by metadata.
 - But also it must leave freedom to users to chose among the two paradigms!
 - As we will see users make a creative use of it

Queries in Hybrid Search

- Any boolean combination of three types of conditions
 - pure semantic:
 - via unique identification of objects/relations
 - e.g. via URIs or unique identifiers
 - keyword-based
 - matching on the whole document
 - keyword-in-context
 - matching keywords only within portion of documents semantically annotated with a specific type or instance

differently from other approaches (e.g. [9]), in HS conditions on metadata and keywords coexist.

Queries in Hybrid Search

- Any boolean combination of three types of conditions
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 - via unique identification of objects/relations
 - e.g. via URIs or unique identifiers
 - keyword-based
 - matching on the whole document
 - keyword-in-context
 - matching keywords only within portion of documents semantically annotated with a specific type or instance
 - e.g. it enables searching for the string "fuel" but only in the context of all the text portions annotated with the concept affected-engine-part [14]

differently from other approaches (e.g. [9]), in HS conditions on metadata and keywords coexist.

nantic Web Tutoria

Example of Hybrid Query

 $\forall x, y, z / (discoloration y) \& (located-on y x) \& (component x)$

Querying Metadata

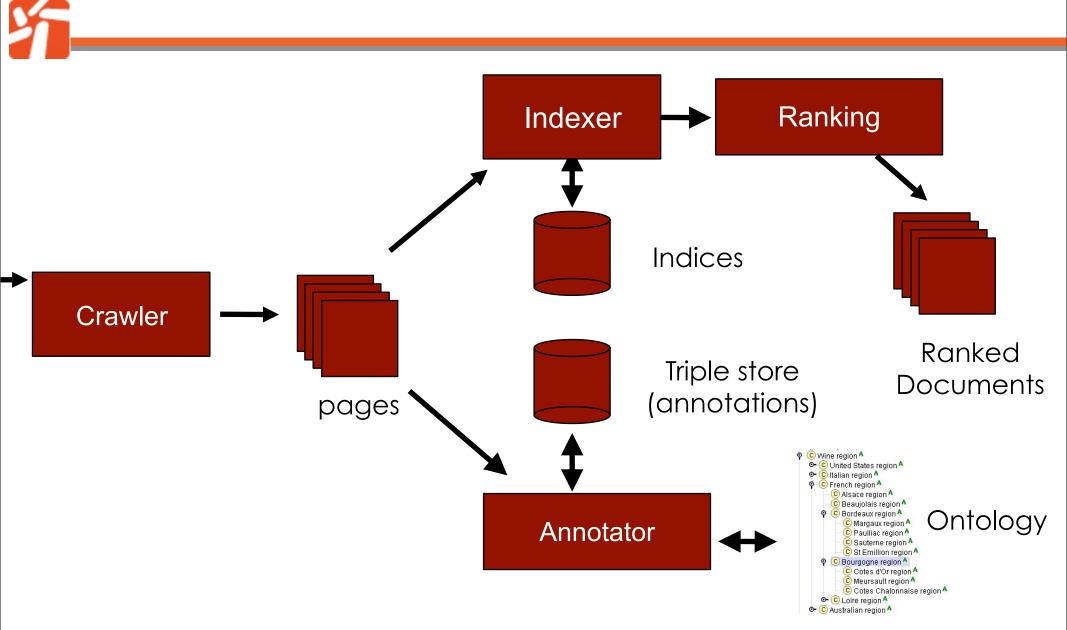
& (provenance-text-contains x "blade")

Keyword in Context Query

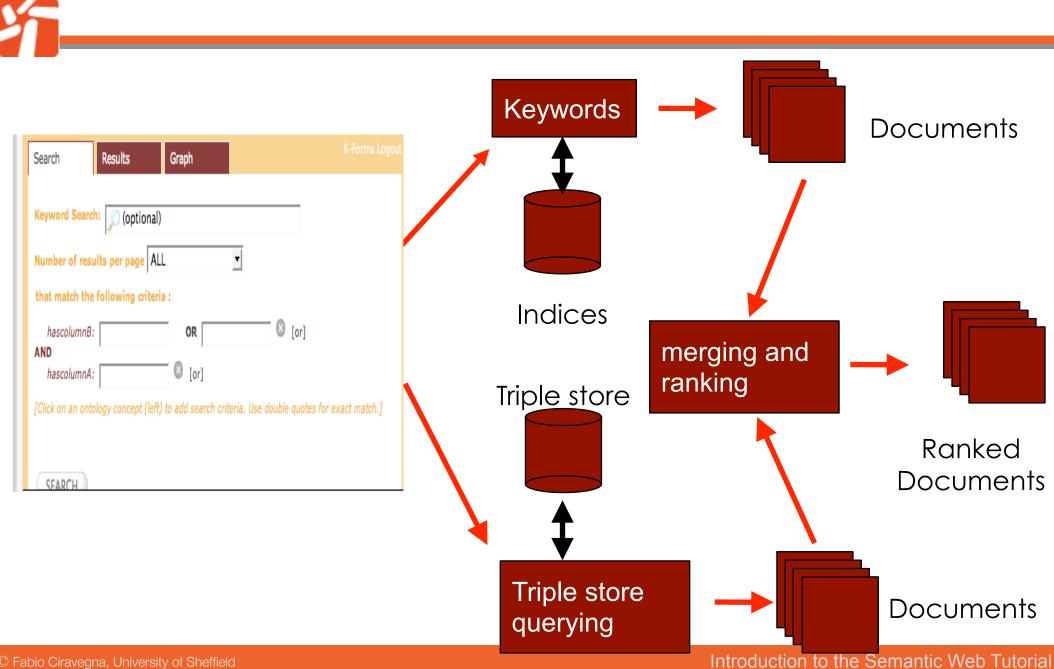
& (contains z "trailing edge") & (document z) & (provenance x z)

Keyword-based Query

K-Search: indexing



K-Search: retrieval



Hybrid Vs Pure Semantic Search

- It is possible to show that Hybrid search can be implemented in all the pure semantic approaches
 - With minimal change
 - Semantic search becomes a subclass of hybrid search

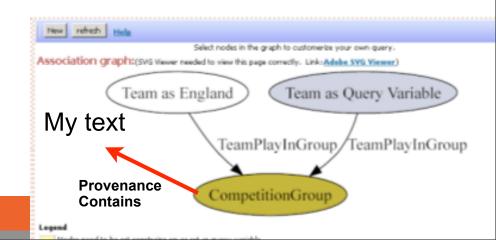
Implementing HS: What Search Strategy?

Semantic Web

- Keyword-based approaches
 - Require translating all the keywords in order to perform the query
 - E.g. SemSearch

lome <u>Knowledge Sources</u> Tools Ontologies	
emantic Search	
This search engine searches relevant data from the back-end semantic data extraction tool ASDI. User can add a subject to narrow down queries by usi project john	
how search summary Refine search	
In search summary Memor search	

- HS implemented by replacing keywords in the query with concepts in the ontology when possible while leaving the rest for pure keyword-based searching
- Keywords in context rather difficult
- View-based approaches
 - Based on querying by building visual graphs
 - E.g. Falcon
 - HS support by adding two arc types
 - document-contains
 - Object description contains



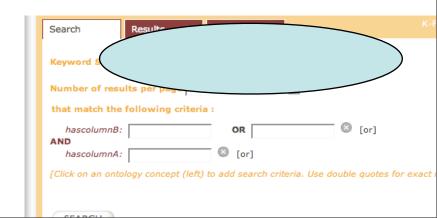
Search Strategy (ctd)

• A natural language approach

- E.g. Aqua
- HS suported by recognising expressions like
 - "and the document contains..."
 - And its description contains
- Form-based approaches
 - HS supported by introducing
 - Keyword Search field
 - Enable keyword Matching on fields

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Available Reports
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ogical Representati					
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Linguistic Triple:		- written - has-author owned-by		esearcher esearcher	- akt - <u>akt</u>
Linguistic Triple: Ontology Triple:	kmi-planet-news-item				- akt - <u>akt</u>



Result Merging

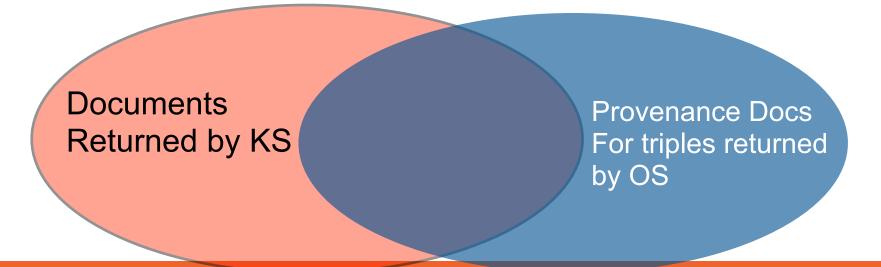


- Merging keyword and semantic results is not straightforward
 - Keyword matching returns an <u>ordered</u> set of URIs of <u>documents</u>
 - a semantic search returns an <u>unordered</u> set of <u>assertions</u> < subj, rel, obj>
- Merging is a different task if:
 - Document Searching
 - Returns documents
 - Knowledge Searching
 - Returns triples

Merging results

- 5
 - Provenance of triples returns document ids for triples (URIs)
 - Document Searching:
 - Provenance URI set is intersected with URIs of documents returned by keywords

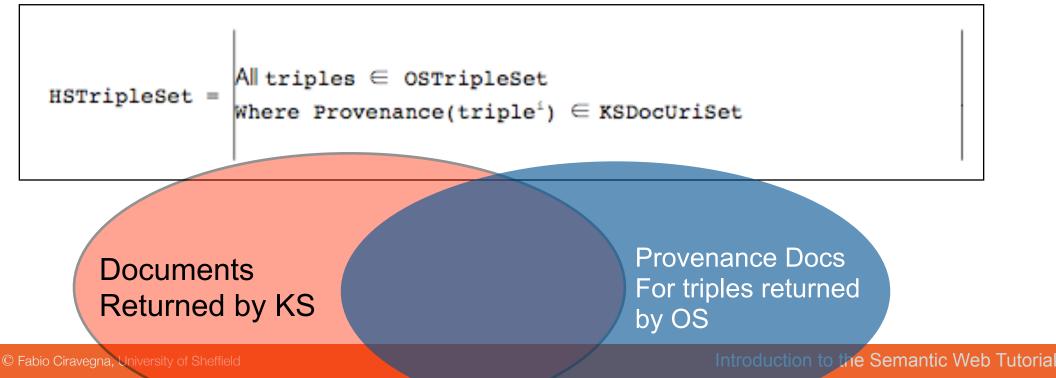




Merging results



- Knowledge Searching
 - Triples returned by semantic search are filtered to remove those whose provenance does not point to any of the documents returned by the keywords



Ranking for Document retrieval

- Effective ranking is extremely important for a positive user experience
- Different ranking methods are possible
 - Document based
 - ability to match the keyword-based query
 - the keywords used in anchor links
 - the document popularity (given by link-based weights)
 - Knowledge Based
 - Presence and quality of metadata

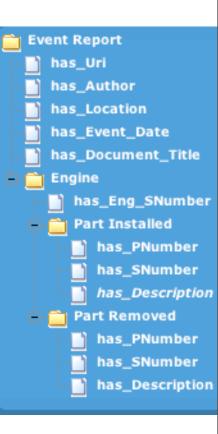


Putting Everything Together

An experience in the aerospace domain

Annotating Documents

- Automatic extraction of information from event report
 - 18,000 documents analysed
 - Mainly Forms implemented in Word
- Metadata generated according to an ontology developed by Aberdeen U
 - Examples manually annotated by users using AktiveMedia
 - Machine Learning + HLT (T-Rex platform) to train the system to annotate
- Automatic extraction of metadata and indexing of documents



Applying information extraction

- AktiveMedia to annotate texts
- TRex system (Jiria et al. 2006) to train and extract
 - <u>http://tyne.shef.ac.uk/t-rex/</u>
- IE captures <u>all</u> the information in tables
 - 99% of the information captured (recall=99)
 - 98% of proposed information is correct (precision=98)

	POS	ACT	CORR	WRONG	MISSED	PREC	REC	F1
airport	120	120	120	0	0	100	100	100
has_airframe_cycles	104	104	104	0	0	100	100	100
has_airframe_hours	104	104	104	0	0	100	100	100
has_author	120	120	120	0	0	100	100	100
has_engine_serial_number	120	120	120	0	0	100	100	100
has_engine_type	120	120	120	0	0	100	100	100
has_event_date	120	120	120	0	0	100	100	100
has_event_report_no	356	358	356	2	0	99	100	100
has_part_description_installed	120	113	111	2	9	98	93	95
has_part_description_removed	120	133	120	13	0	90	100	95
has_part_number_installed	120	113	111	2	9	98	93	95
has_part_number_removed	120	133	119	14	1	89	99	94
TOTALavegna, University of Sheffield	1644	1658	1625	33	19ntro	ducti 98 to tl	ne S 99 1ant	ic V 98) Tu

oria



- Form-based implementation of hybrid search initially created for Jet Engine Designers
- It enables
 - Document querying

k-now.co.uk

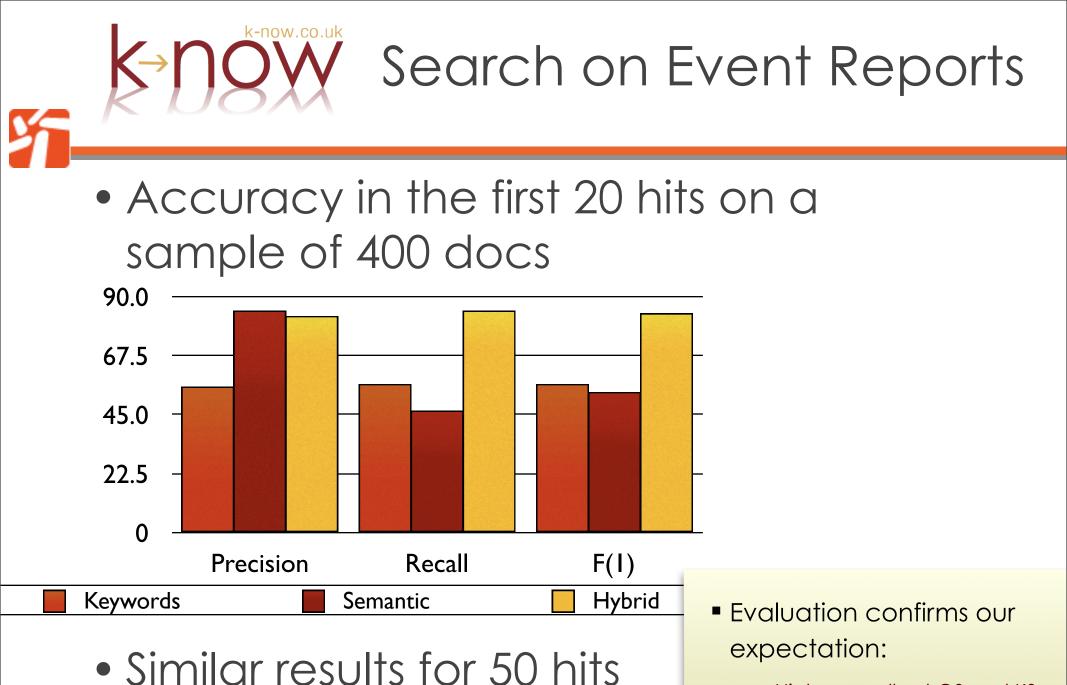
- Knowledge querying
 - Including quantification of unstructured information

			graph2	
Available Reports	Search Results Keyword Search: Decision Results Keyword Search: Decision Results per page ALL that match the following criteria hascolumnB: blah AND hascolumnA: blahblah [Click on an ontology concept (left)]	Pelererd Service Event Service Event Oate Poate Poate	The University of Sheffield. Image: Sheffield. <	Pie Chart
hascolumnB hascolumnA	match.]	Part Number Serial Number Description Component Orart Number Sarial Number	Event Report Data engine name here Rolls-Royce place here	7% ● fcoc 7% ● front combustion outer case 23% ● "fan cowl door, J/h" 15%) protection box 7% ● vigv 7% ● thrust reverser rh c-duct 7% de assembly 15% ● "fan cowl door, rh" 7%
© Fabio Ciravegna, University	of Sheffield	- O Description	WB612 LN184 Event Date: 09-Nov-01 Engine S/N: 51127 Flight Regime: Hazard Type: Aircraft Regn: 97-SYB Installed Posn: Right Location: IN No Hazard Airfame Hours: 5535 Engine TSN/CSN: Installed Posn: Int Type: Non-Dertained	tion to the Semantic Web Tutoria



K-Search evaluation

- We have performed 2 types of technology evaluations using K-Search:
 - in vitro:
 - Effectiveness of annotation and query strategy with respect to standard KS and OS
 - in vivo: testing the system with real users
 - 32 users Rolls-Royce engineers
 - Evaluation enables verifying suitability for use in a real environment



- Higher recall wrt OS and KS
- Higher precision wrt KS
- Slightly lower precision wrt OS



Final User Evaluation

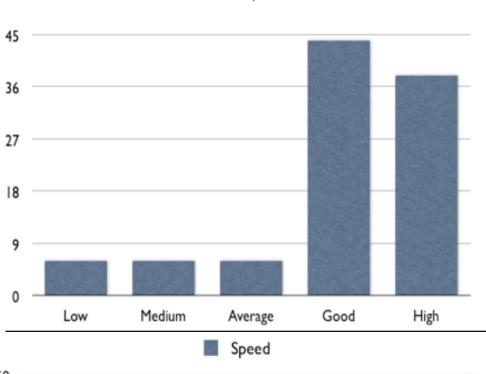
- Goal: verifying suitability for use in a real environment
 - 32 users Rolls-Royce engineers from different parts of the company
 - 90 minutes of test
 - Short introduction
 - 3 monitored tasks
 - One given (including solution)
 - One given (no solution)
 - One free task
 - Availability of system on intranet for the following period
- Evaluation: video recording, interview + log analysis



Evaluation Questions

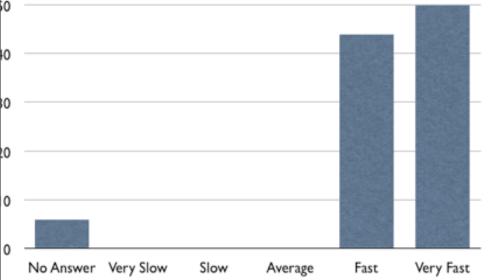
- Do user understand the hybrid paradigm?
- Are they able to search using HS?
- Do they actually use HS when confronted with a real searching task?
- Would the users be willing to use the system for their everyday work?

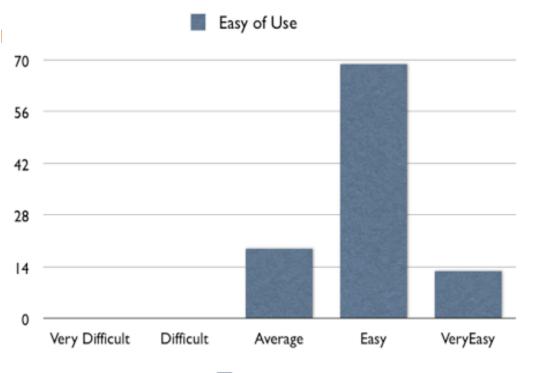
Liked by users?



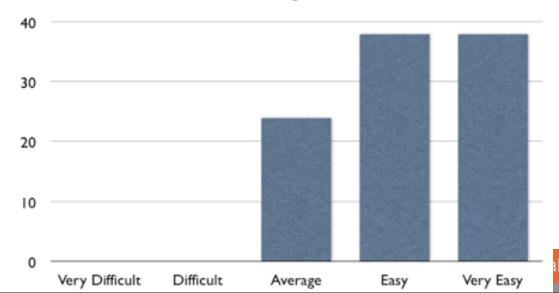
k-now.co.uk

Accuracy





Learning to Use





Liked by Users?

- Finalist of Rolls-Royce Director's Creativity Award 2007
 - Voted by senior employes for its innovation potential





Liked by Users?

- Support to the design of new jet engine
 - Porting to 9 Information Sources
 - 2008-2009
 - Carried out by:
 - 50% University
 - 50% k-now Itd (university spinout-company)
- Funds requested to UK Government for use of K-Tools for use in manufacturing





Conclusions

- Document annotation can be performed at different levels
 - Ontology-based, braindump, document enrichment
- User centred automated ontology-based
 annotation
 - For trusted self contained documents (e.g. KM)
 - AktiveMedia
- Automated means of capturing knowledge
 - Several Tasks

Conclusions



- Sharing and Reuse
 - We have seen
 - Document Enrichment
 - Semantic Search
 - Different paradigms for search

Future Work & Challenges

- Multidisciplinary research for annotation
 - NLP has strong role, but complemented with other disciplines
 - SE, ML, II, SWS, HCI
- Annotation
 - Beyond the division between user centred and unsupervised
 - Strong HCI strategies
 - Validation of results across documents
 - » How can you validate 2M triples produced by large scale annotation?

Future Work & Challenges (2)

- Modelling:
 - How modelling uncertainty?
 - Knowledge is dynamic. How do you model that?
- HCI
 - Information presentation (document annotation)
 - Intrusivity:
 - How to avoid annoying users with too many annotations
 - Trust
 - Who do users trust?
 - » Tracing preferred sources
 - Where does the information come from?
- Scalability
 - Large scale indexing systems
 - Millions of pages (not billions!)

Conclusions and Future Work

- The Semantic WEB offers <u>potentially</u> key technologies to the development of future knowledge Management and the Web
 - More Web than Semantics, but:
 - A little semantics goes a long way (J. Hendler)
- The potential must be exploited addressing <u>real</u> <u>world</u> requirements
 - Rather than in principle Al-oriented requirements (e.g. closed world, small scale, etc.)
- Strong application pull can be obtained
 - Do not sell slogans, sell ideas and applications!









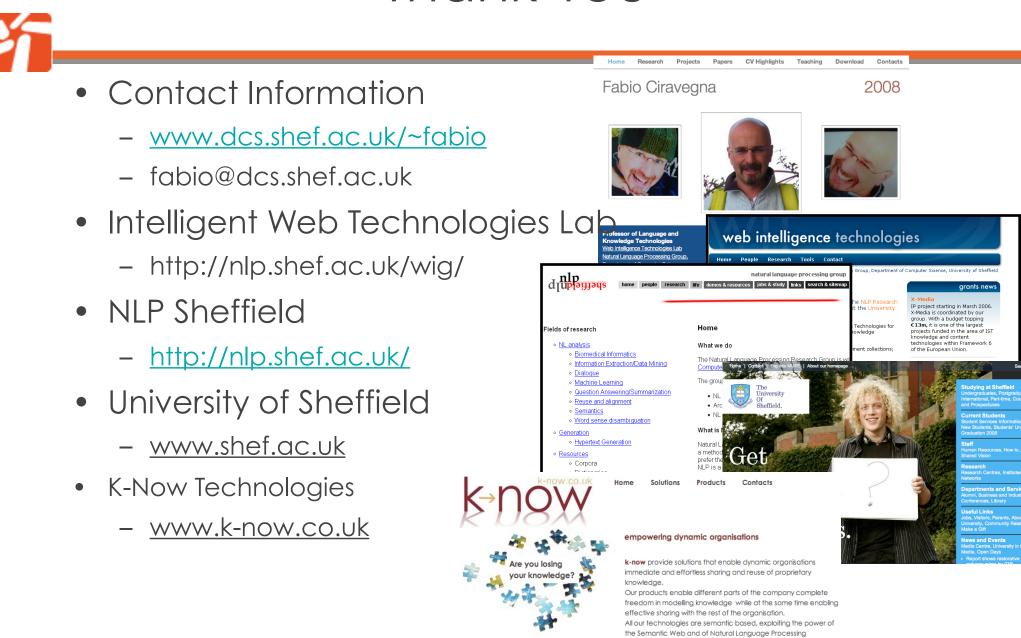


A final thought

- 5
- These technologies allow easy collection of and access to a *very* large amount of information/ knowledge
- Are we:
 - Preparing for a better Web/better world?
 - Preparing for a world with no privacy?
 - Big brother
 - Spam
 - Identity theft (e.g. Garlik)
 - Just adding hay to the haystack while searching for a needle?
 - Drowning in triples while trying to avoid drowning in texts?

The Karen Spark-Jones slide

Thank You



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