



Semantic Web Technologies for Capturing, Sharing and Reusing Knowledge

- aka Annotating and
 Searching the Semantic Web
- aka: HLT and ML for the SW

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image ® Rolls-Royce

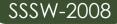
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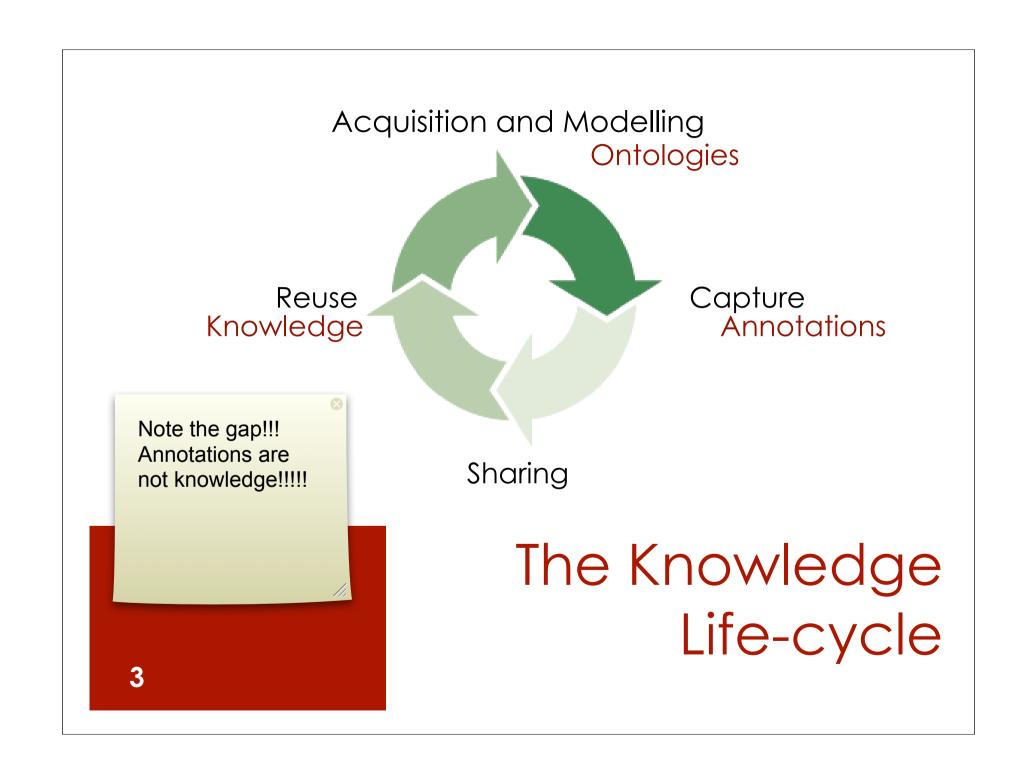
■ These slides were presented during the The Sixth Summer School on Ontological Engineering and the Semantic Web (SSSW'08), July 6-12, 2008. Cercedilla(Spain)

http://kmi.open.ac.uk/events/sssw08/

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Why manage knowledge?

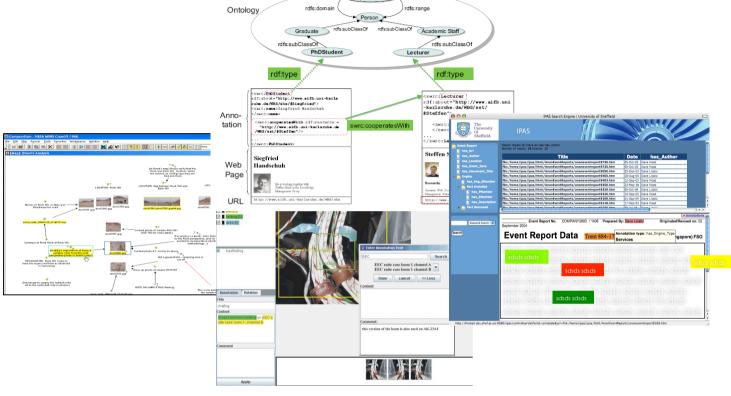
- To enable easy <u>timely and effective</u> reuse
 - We need: to enable sharing
 - Requirements: easy and effective sharing
- To enable sharing
 - we need to: capture knowledge
 - Desiderata:
 - Easy capture (do not get in the way of the user's work!)
 - Comprehensive capture (do not miss important 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 + process support
- You have already seen:
 - Knowledge Acquisition and Modelling
 - Ontology Engineering





Requirements for Knowledge Capture

- issues in knowledge capture:
 - capture: what and what for?

Knowledge Capture

- Collecting and aggregating multimedia knowledge to make it available for
 - sharing and reuse
 - From document management to knowledge management
 - for integration

In ontological terms knowledge capture consists in capturing instances!

- Approaches
 - at source: helping people capturing knowledge when produced
- On legacy documents, pictures, data:
 - Annotation services



Requirements for KC: Cross media

Evidence is often distributed in different media;

Knowledge in one medium does not carry the full

evidence

Battery Exchange Program iBook G4 and PowerBook G4

Apple has determined that certain lithium-ion batteries containing cells manufactured by Sony Corporation of Japan pose a safety risk that may result in overheating under rare circumstances.

The affected batteries were sold world. 2003 through August 2006 for use wit notebook computers: 12-inch iBook G. PowerBook G4 and 15-inch PowerBook

Apple is voluntarily recalling the affects has initiated a worldwide exchange progligible customers with a new replacer charge. This program is being conduct, with the U.S. Consumer Product Safety (CPSC) and other international safety a.

Identifying your battery

Please use the chart below to identify the and serial numbers that apply to your in PowerBook. If the first 5 digits of your is serial number fall within the noted range replacement battery immediately.

To view the model and serial numbers labeled on the bottom of the battery, you must remove the battery from the computer. The battery serial number is printed in black or dark grey lettering beneath a barcode. See photos below.

this case is no longer valid because we have introduced Service Note 3445 which requires replacement of component







Compound Documents & CM

From Deliverable D8.2

- Typical data objects (text, image, raw)
- 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

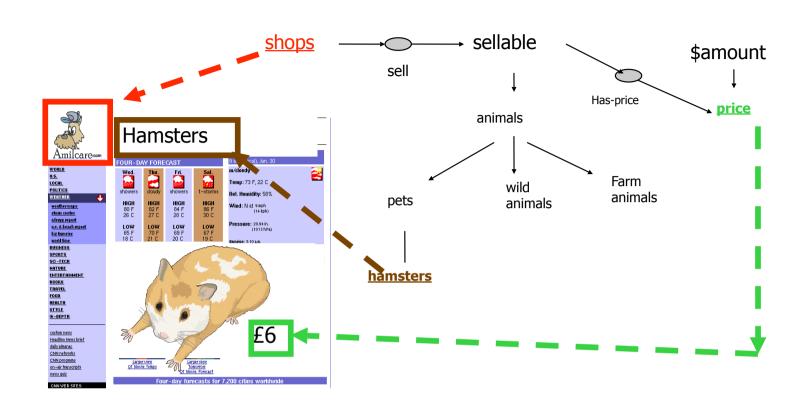












SW for Knowledge Capture

- user centred methodologies and tools for text and image annotation
- automatic methodologies and tools for text annotation

Semantic Web for Knowledge Capture

Aims:

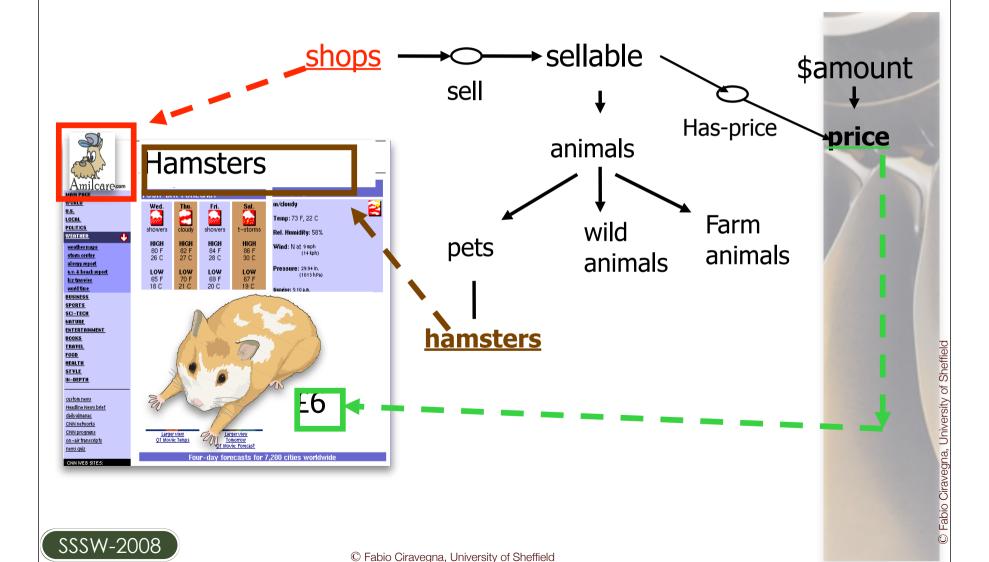
- To capture knowledge within and across media in a rich, semantically-oriented way
- Outcome of capture technologies is a semantic representation of the content (conceptualisation) to be used for knowledge management purposes
- Enrichment of multimedia documents with layers of manually or automatically generated annotation is the main medium of associating conceptualisations to resources

Ontology-based annotations

- Marking up contained information
 - Portions of documents associated to objects in ontology
 - Allows:
 - Ontology-driven processing
 - Services based on ontology will be able to use information
 - Ontomat/CREAM (Staab et al 2001)
 - Melita (Ciravegna et al. 2002)
 - SemTag and Seeker (Dill et al. 2003)
 - ...and many others...

Ontology-based Annotation

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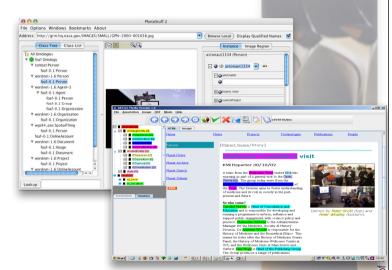


Input &Output

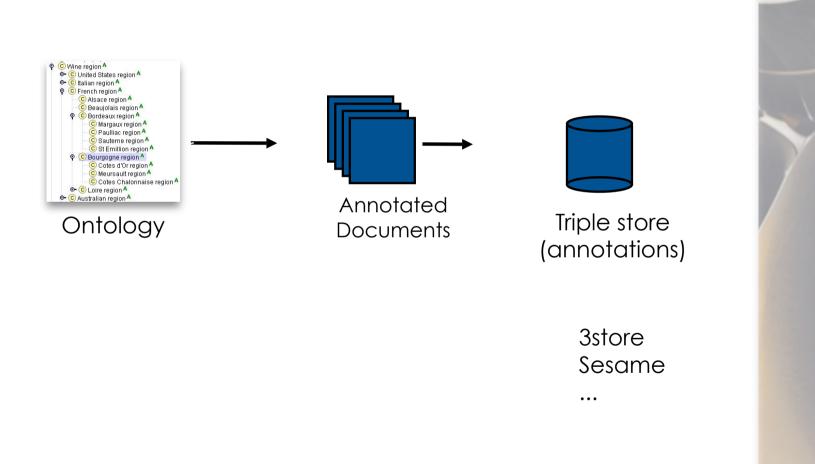
- Input to the KC technologies
 - Ontologies (MMO, domain ontology),
 - Background knowledge (gazetteers, etc.)
 - Normalised document representation
 - Medium to extract from (text, images, data, videos,...)
- Output
 - Evidence represented in terms of conceptual information
 - Evidence used by other modules as background conceptual knowledge, i.e. pre-existing knowledge
 - Evidence in the form of uncertain output

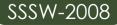
Ontology-based Annotation

- The way to annotate pages is to:
 - Select an ontology
 - Define statements to represent meta-data about the document
- Manual Annotation
 - Annotation can be performed by:
 - Domain expert
- 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)



Ontology-based Annotation

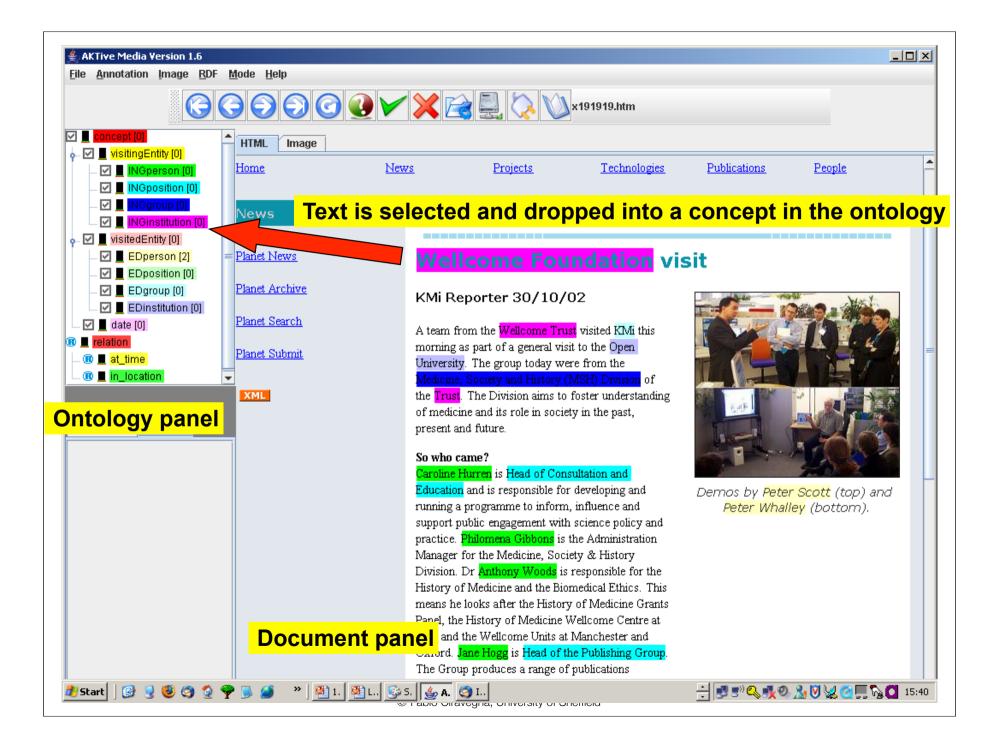




AktiveMedia

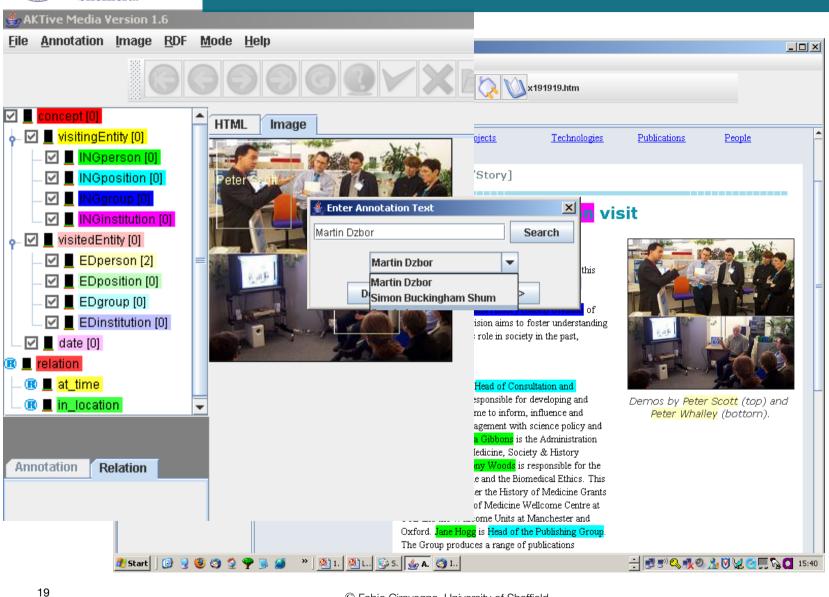
- Enables semi-automatic annotation across texts and images
- The interface enables
 - HTML editing
 - Annotation of documents in RDF based on an OWL ontology
- Types of annotations
 - Concepts / Relations
- SW: Annotation:
 - Selection of concept/relation and highlighting of text is the way in which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

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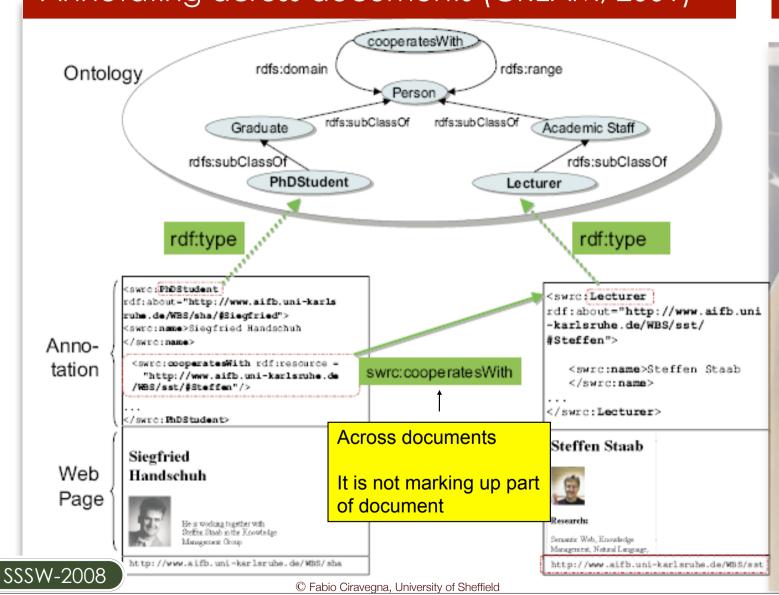
Contextual Annotation of Images and Text



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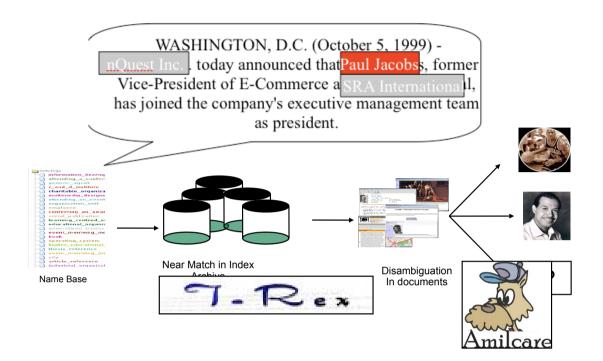
Annotating across documents (CREAM, 2001)



Marking up Provenance

- COMM A Core Ontology for Multimedia based on
 http://comm.semanticweb.org/
 - the MPEG-7 standard
 - the DOLCE foundational ontology.

```
<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>
```



Automating Annotation

Annotation Engines

- 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
 - 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

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Information Extraction from Text

- 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, o

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

Company: nQuest Inc.

Date: today

InPerson: Paul Jacobs

InRole: president

Company: SRA International

OutPerson: Paul Jacobs

OutRole: Vice-President of E-Commerce,

Named Entities

Event Recognition

Growing complexity

SSSW-2008

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Classic Tasks

- Information Extraction from Text:
 - 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

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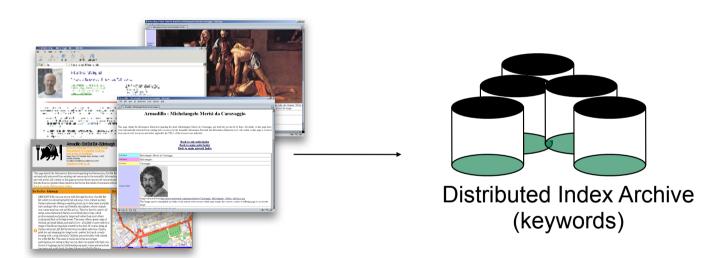
Large Scale NER&C

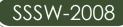
- For large scale (some hundred millions pages) smarter infrastructure is needed
 - Search engine-like indexing infrastructure
 - Faster processing (less processing)
 - Two cases:
 - Recognition of known terms (and their variations)
 - See also information integration
 - Discovery of new names



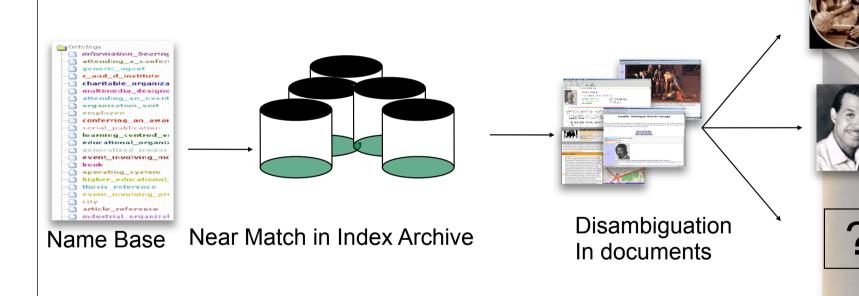
Large Scale NER: Indexing

Document Indexing as in Search Engines





Known Name Recognition

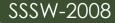


S. Dill, N. Eiron, et al: SemTag and Seeker: Bootstrapping the semantic web via automated semantic annotation. WWW'03

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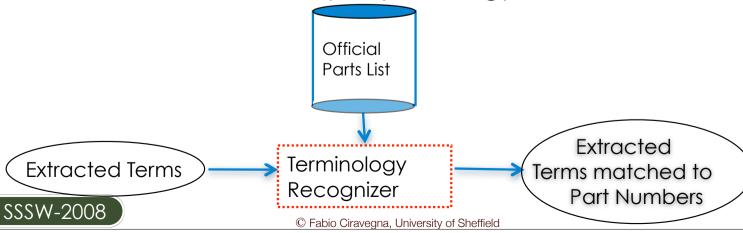
Discovery of New Names

- 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

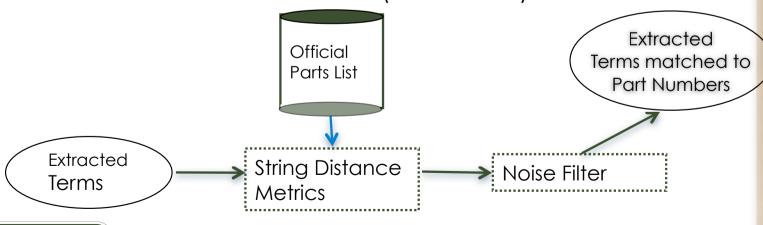
- NER is one example of term recognition
- More useful in technical domains is terminology recognition
 - The task of assigning a URI to a technical description
 - i.e. mapping a natural language description to the official company ontology



Terminology Recognition

- Possible approaches
 - Linguistic approaches
 - Based on linguistic analysis of terms (Gaizauskas et al 2003)
 - Statistical approaches
 - Based on frequency analysis and detection
 - Other approaches

■ Distance metrics based (Butters 2007)



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Table Field Extraction

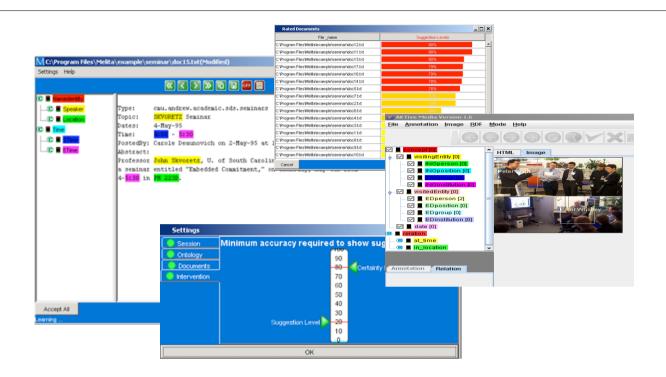
- Tables are an essential part of many documents
 - Most information is represented in tables
- Tables can be represented as forms to fill
 - Semantics is fixed
 - Wrapper writing or wrapper induction (Kushmerick 1997)
- Tables can be created ad hoc in documents (e.g. Word docs)
 - Semantics is unclear
 - 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

- 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

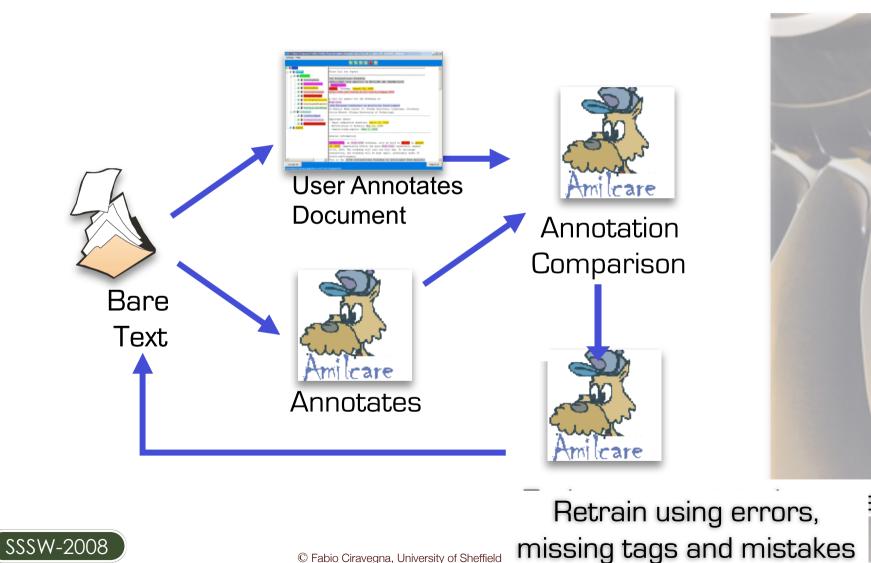
A list of tools for automatic annotation

- 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/



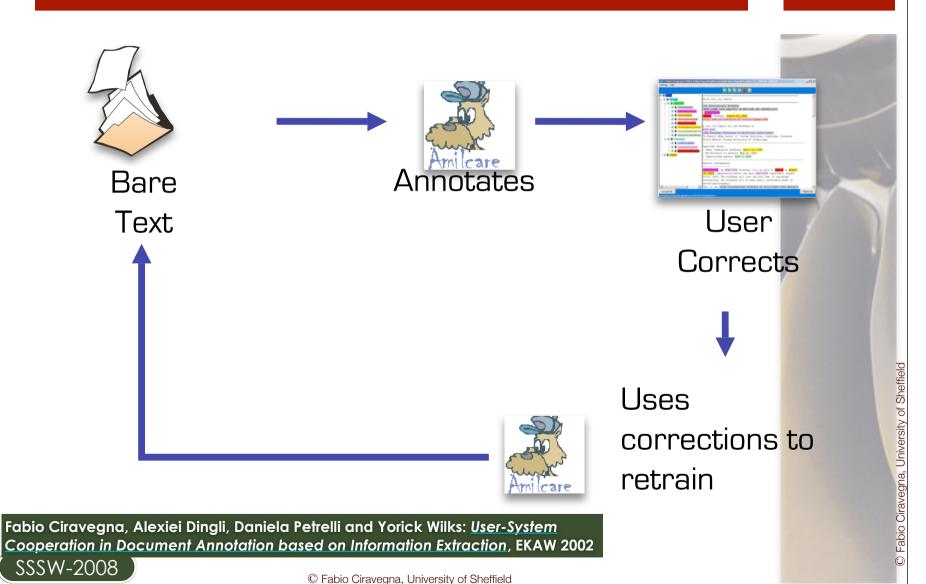
Using IE to Support Manual Annotation

Using IE to support annotation: step 1

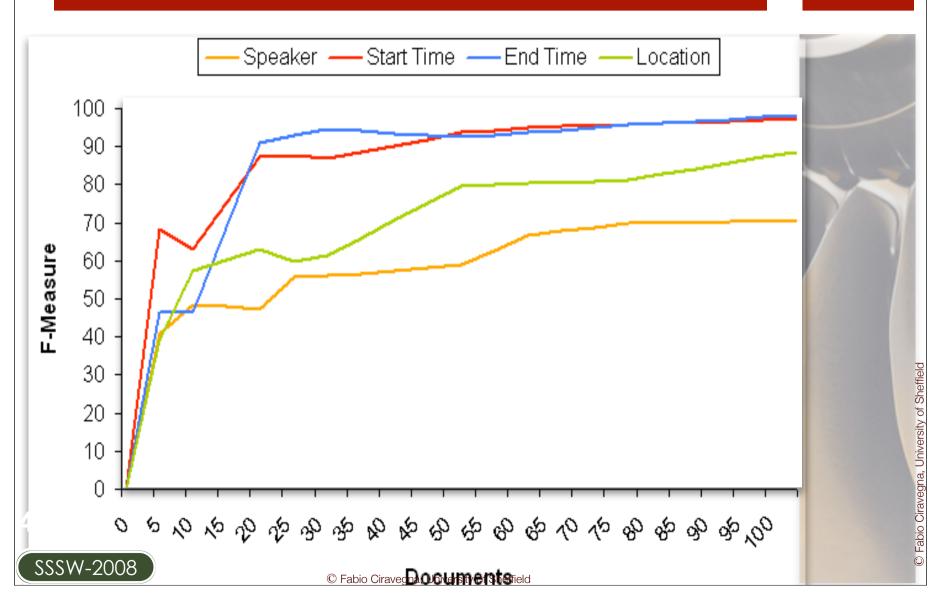


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Using IE to support annotation: step 2

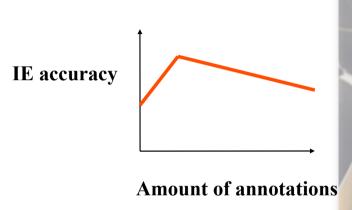


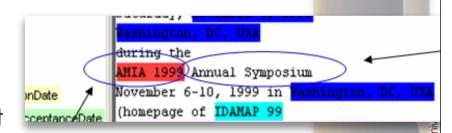
Learning curve



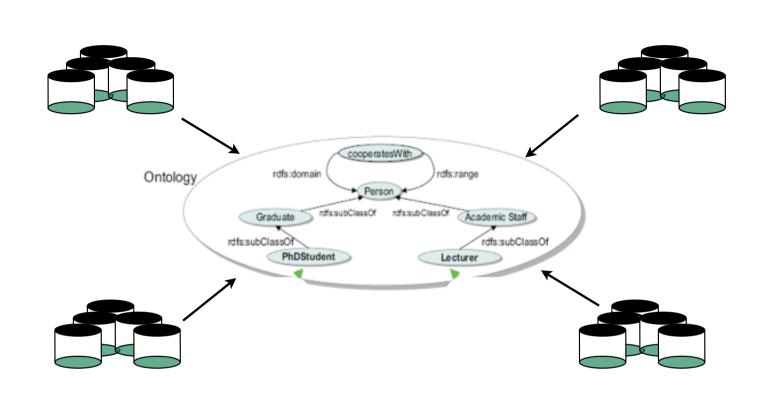
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- University of Karlsruhe experiments
 - -80% annotation time
 - +100 interannotator agreement
 - Is this positive?
- Outstanding issue:
 - Impact on annotators of suggestions topping 85% accuracy?
 - Annotation needs to be precise and consistent
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





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Information Integration

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)

- Statistical matching
- Application specific
 - Rules

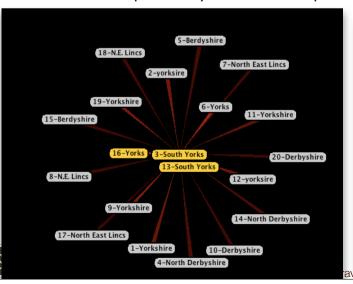
Using Web bias

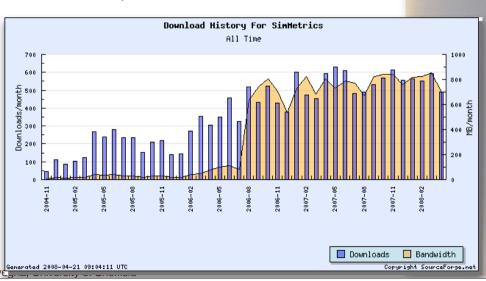
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SimMetrics

- Library of distance metrics released as open source
 - http://sourceforge.net/projects/simmetrics/
 - >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



Sources

The Marine Society Registers

Prerogative Court of Canterbury Wills

The Westminster Historical Database

The Proceedings of the Old Bailey

Fire Insurance
Policies

AHDS Deposits

St. Martin's
Settlement Exams
Index
WESTCAT

Metropolitan
London in the 1690s

IHR

Collage image databse
Guildhall Library

Selected Criminal Records PRO Harben's Dictionary of London

John Strype's "Survey..."

http://www.motco.com



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KING'S Committee of the E

Oyer and Terminer, and Goal-Delicery of Newgete, held for London and COUNTY of Middlefee, at Yoffice Hall in the On Westerfee, Terrifae, and Friday, being the 16th, 17th, and 18th January, in the Ninth Yore of His MAJESTY's Ro

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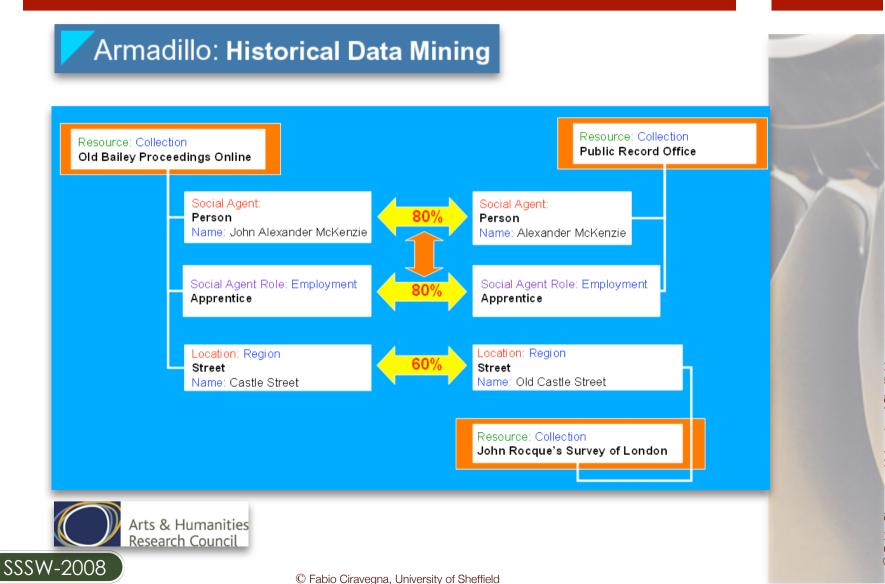


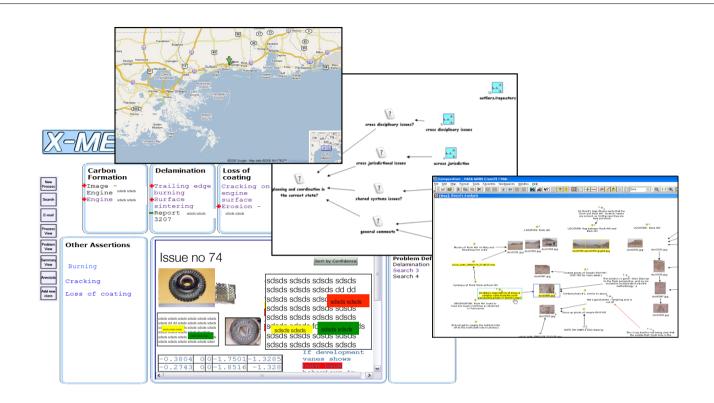


House of Lords
Journals
BOPCRIS

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

Information Integration



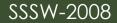


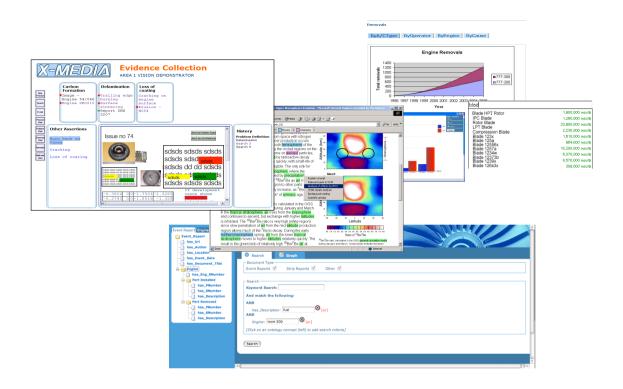
Knowledge Sharing and Reuse

- issues in knowledge sharing
- approaches and novel methods to searching, sharing and reuse knowledge

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





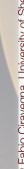
SW for Knowledge Sharing and Reuse

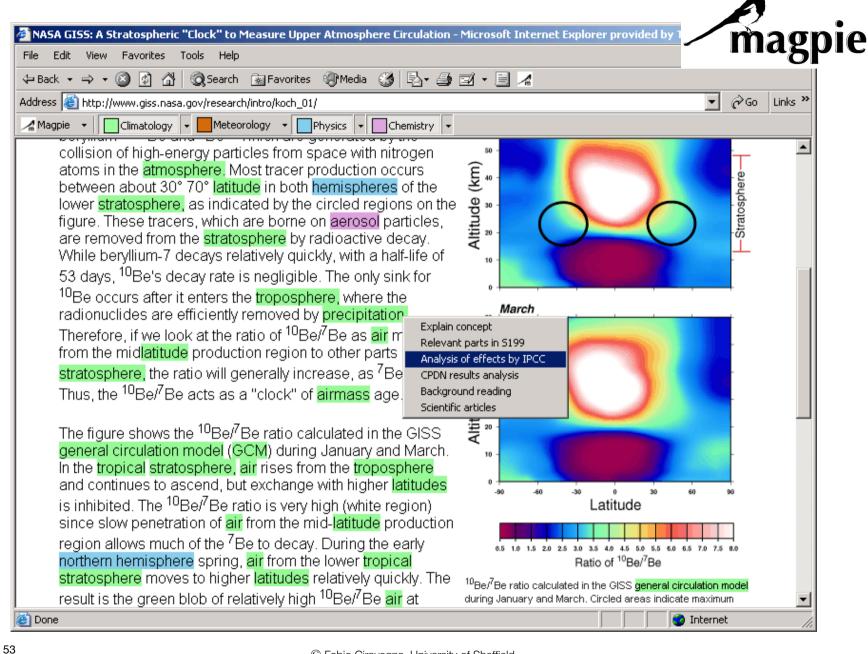
Sharing and Reuse via SW

- 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

Keywords Based Searching

- KS effectiveness is often affected by 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)

Ontology-based Search

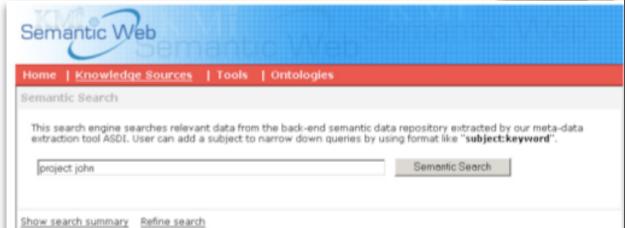
- Searching metadata rather than texts or images
 - Ontology enables reasoning
 - More flexible than searching using traditional methods
- Searching to...
 - Retrieve documents (images/texts/videos/data)
 - As replacement of traditional document management systems
 - Retrieve information/knowledge
 - Querying the knowledge (e.g. the triple store)

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
 - All the keywords are mapped to Semantic Concepts
 - Requirements: feedback on generated query
 - Issues:
 - User lost for words
 - What is covered by the ontology?
- E.g. SemSearch





Semantic Search Approaches (2)

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

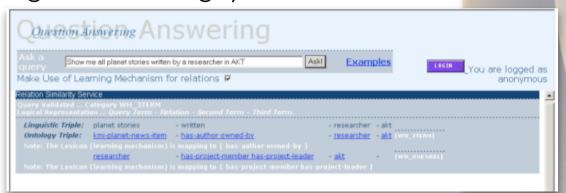
querying

■ E.g. Falcon



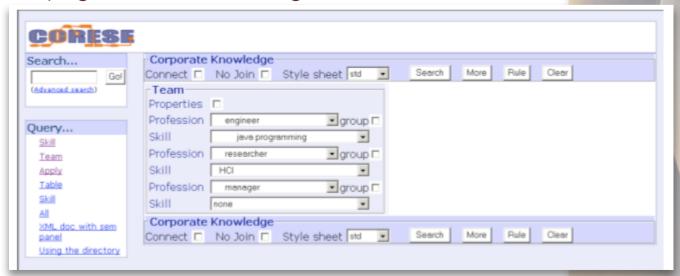
Semantic Search Approaches (3)

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



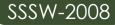
Semantic Search Approaches (4)

- 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



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

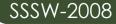


An Experiment on Jet Engine Event Reports

- 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 all the events associated with damage to acous- tic liners following bird strike"
 - How many topics can we model with Information Extraction?
 - 21 topics/ 14 topics partially or not covered by IEbased 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 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 type conditions

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

- 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
 - 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]

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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

Implementing HS: Indexing

- Documents are indexed using a standard keyword-based engine such as SolR
- Facts (e.g. extracted by an IE system) are stored in a Knowledge Base
 - e.g. a triple store like Sesame2 in the form of RDF triples.
- Provenance of facts recorded
 - E.g. As triples connecting
 - the facts' URIs and those of the document of origin
 - the facts' URIs and the original strings used in the documents

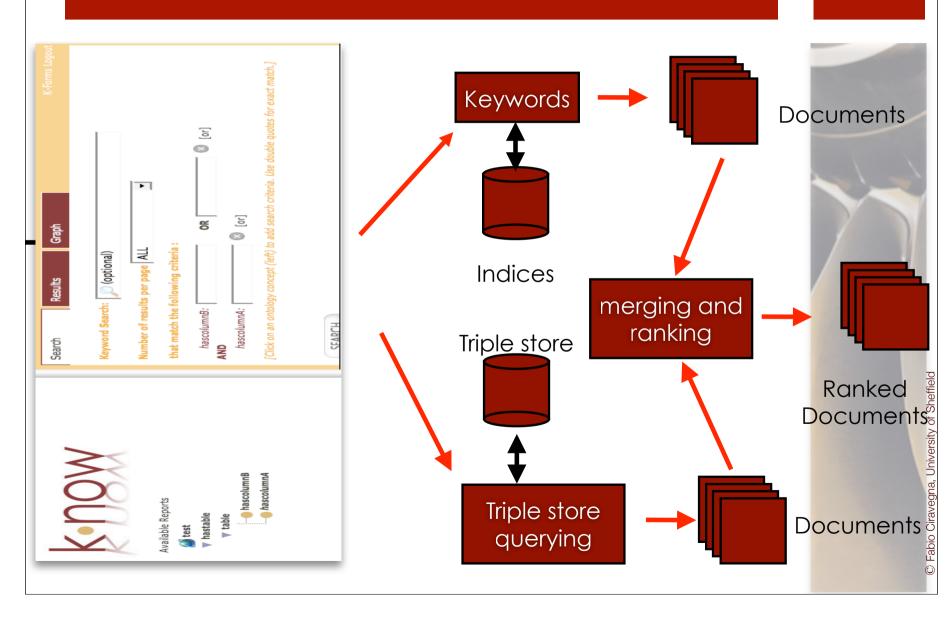
K-Search: indexing Ranking Indexer Indices Crawler Ranked Triple store **Documents** (annotations) pages P © Wine region A ◆ © United States region ♣ ● © Italian region A P © French region A C Alsace region A © Beaujolais region A Ontology P © Bordeaux region A 🕒 Margaux region 🗛 Annotator Paulliac region A Sauterne region A C St Emillion region A P © Bourgogne region A Cotes d'Or region 🗛 © Meursault region 🗛 Cotes Chalonnaise region • ● C Loire region A

Implementing HS: Retrieval

- Query is parsed and the different components (keywords, keywords-in-context and metadata) identified
 - keyword matches → traditional information retrieval system
 - metadata searches
 - Translated into a query language like SPARQL
 - Sent to a triple store
 - keywords-in-context queries
 - matched with provenance of annotations in documents
 - E.g. Using SPARQL and a triple store
- Finally, results are merged, ranked and displayed

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K-Search: retrieval



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

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I won't mention

ranking here

Merging results

- Provenance of triples returns document ids for triples (URIs)
 - Document Searching:
 - Provenance URI set is intersected with URIs of documents returned by keywords
 - HybridSearchUriSet= KSDocUriSet ∩ OSDocUriSet

Documents Returned by KS

Provenance Docs For triples returned by OS Ciravegna University of Sheffie

I won't mention

Merging results

- Provenance of triples returns document ids for triples (URIs)
 - Knowledge Searching

■ Triples returned by semantic search are filtered remove those whose provenance does not point to

any of the documents returned by the keywords

```
HSTripleSet = All triples = OSTripleSet
Where Provenance(triple1) = KSDocUriSet
```

Documents Returned by KS Provenance Docs For triples returned by OS

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

Expected effect of HS: Document Searching

- With respect to OS
 - Recall expected to increase
 - Use of keywords where metadata is missing enables to answer otherwise impossible queries
 - Precision may suffer because of polysemy
- With respect to KS
 - Precision and recall expected to increase
 - Ambiguity and synonymity are dealt with by semantic search when available
 - Higher recall and precision
 - As keywords are combined with metadata in the same query, the context given by the available metadata helps in disambiguating keywords as well
 - higher precision



Expected effect of HS: Knowledge Searching

- With respect to OS
 - Precision increased
 - Use of keywords where metadata is missing enables more precise queries
 - although less precise than the ideal ones



- Recall increased
 - Use of keywords where metadata is missing enables to answer otherwise impossible queries
- Precision may suffer because of polysemy
- With respect to KS
 - KS does not cover Knowledge Searching

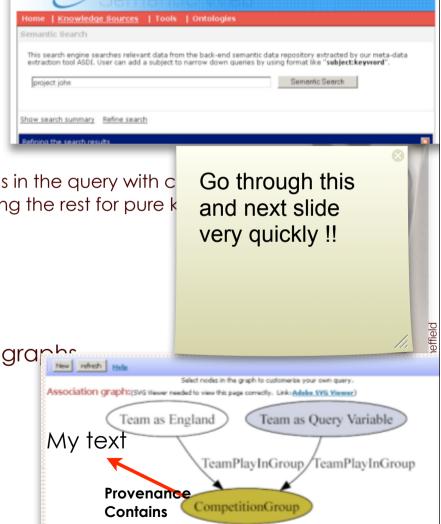
Next slide:
We have
implemented a
version to confirm
our expectation

Implementing HS: What Search Strategy?

Semantic Web

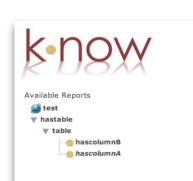
Legend

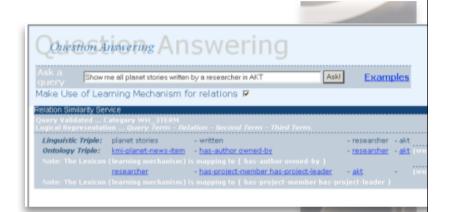
- Keyword-based approaches
 - Require translating all the keywords in order to perform the query
 - E.g. SemSearch
 - HS implemented by replacing keywords in the query with c the ontology when possible while leaving the rest for pure k based searching
 - Keywords in context rather difficult
- View-based approaches
 - Based on querying by building visual graphs
 - E.g. Falcon
 - HS support by adding two arc types
 - document-contains
 - Object description contains



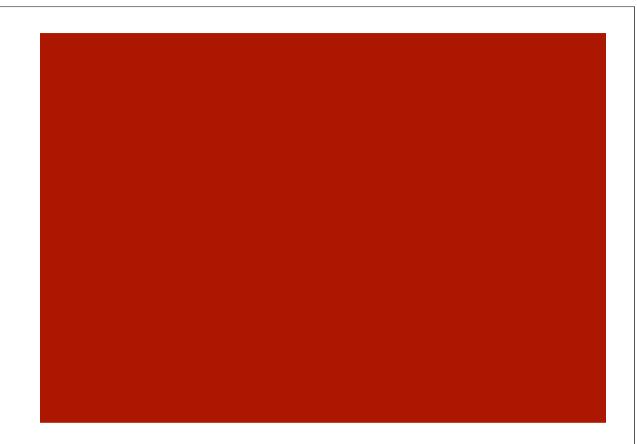
Search Strategy (ctd)

- 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
 - KeywordSearch field
 - Enable keywordMatching on fields





 Form-based implementation of hybrid search initially created for Jet Engine Designers



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

IE unable to cover all the ontology with sufficient accuracy

Applying information extraction

- AktiveMedia to annotate texts
- TRex system (Jiria et al. 2006) to train and extract
 - http://tyne.shef.ac.uk/t-rex/
- 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
TOTAL	1644	1658	1625	33	19	98	99	98



K-Search

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

Including quantification of unstructured information



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

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In vitro evaluation

- 21 topics of search, discussed with users, 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"

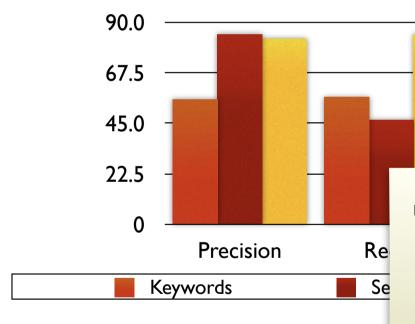
• Queries:

- "what events caused during maintenance in 2003 were due to control units?"
- Translated into a set of queries in KS, OS and HS

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K-Search on Event Reports

 Accuracy in the first 20 hits on a sample of 400 docs



Similar results for 50 hits

- Evaluation confirms our expectation:
 - Higher recall wrt OS and KS
 - Higher precision wrt KS
 - Slightly lower precision wrt OS

Final User Evaluation

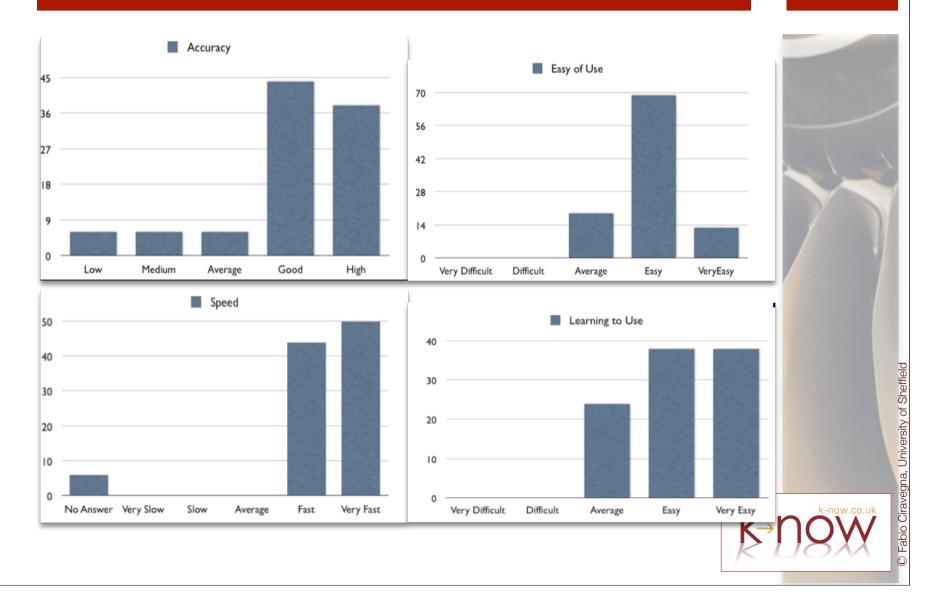
- Goal: verifying suitability for use in a real environment
 - 32 users Rolls-Royce engineers from different parts of the company
 - 90 minutes of test
 - Short introduction
 - 3 monitored tasks
 - One given (including solution)
 - One given (no solution)
 - One free task
 - Availability of system on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

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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 the users?



Liked by Users?

- Finalist of Rolls-Royce Director's Creativity Award
 2007
 - Voted by 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 ltd (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



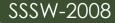
Future Work & Challenges

- Multidisciplinary research for automation
 - 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?

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Future Work & Challenges (2)

- 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> world 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

- These technologies allow easy collection of
 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

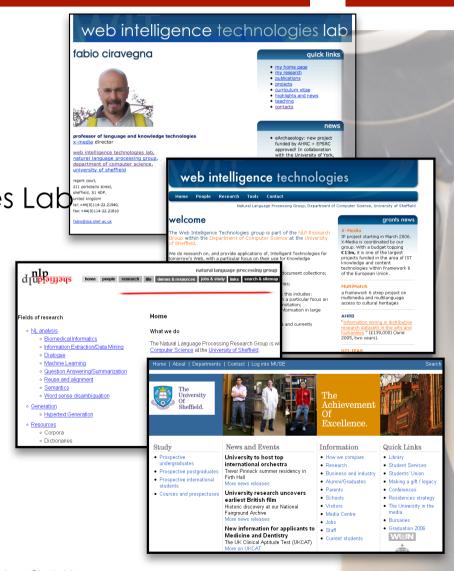
The Karen Spark-Jones slide

- Identity theft
- Just adding hay to the haystack while searching for a needle?
 - Drowning in triples while trying to avoid drowning in

texts?

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- Contact Information
 - www.dcs.shef.ac.uk/~fabio
 - fabio@dcs.shef.ac.uk
- Intelligent Web Technologies Lab
 - http://nlp.shef.ac.uk/wig/
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