

# Natural Language Processing for Information Access

Horacio Saggion

Department of Computer Science
University of Sheffield
England, United Kingdom
saggion@dcs.shef.ac.uk





#### Overview of the course

- NLP technology and tools (Day 1 & 2)
- Question Answering (Day 3 & 4)
- Text Summarization (Day 4 & 5)





#### **Outline**

- NLP for Information Access
- Information Retrieval
- Information Extraction
- Text Summarization
- Question Answering
- Cubreporter: a case study
- Other applications
- GATE tools for NLP

- Components
  - tokenisation
  - sentence splitting
  - part of speech tagging
  - named entity recognition
  - morphological analysis
  - parsing
- Demonstrations





# Information Retrieval (Salton'88)

- Given a document collection and a user information need
- Produces lists of documents matching the information need
- Information needs can be expressed as sets of keywords
- Documents are pre-processed in order to produce term indexes which contain information about where each term occurs in the collection - decisions have to be taken with regards to the definition of term





#### **Information Retrieval**

- Needs a method to measure the similarity between documents and queries
- The user has to read the documents in order to find the desired information
- IR can be applied to any domain
- Text Retrieval Conferences (since 1992) contributed to system development and evaluation (http://trec.nist.gov)





# Information Extraction (Grishman'97)

- Pulls facts from the document collection
- Based on the idea of scenario template
  - some domains can be represented in the form of one or more templates
  - templates contain slots representing semantic information
  - IE instantiates the slots with values
- IE is domain dependent a template has to be defined
- Message Understanding Conferences 1987-1997 fuelled the IE field and made possible advances in techniques such as Named Entity Recognition
- From 2000 the Automatic Content Extraction (ACE) Programme





#### Information Extraction

ALGIERS, May 22 (DFP) - At least 538 people were killed and 4,638 injured when a powerful earthquake struck northern Algeria late Wednesday, according to the latest official toll, with the number of casualties set to rise further ... The epicentre of the guake which measured 5.2 on the Richter scale) was located at Thenia, about 60 kilometres (40 miles) east of Algiers, ...

DATE

21/05/2003

DEATH

538

INJURED

4,638

EPICENTER

Thenia, Algeria

INTENSITY

5.2, Ritcher





#### **Information Extraction**

- Template can be used to populate a data base
- Template can be used to generate a short summary of the input text
  - A 5.3 intensity earthquake in Algeria killed more than 500 people.
- Data base can be used to perform reasoning
  - What Algerian earthquake killed more people?





#### **Information Extraction Tasks**

- Named Entity recognition (NE)
  - Finds and classifies names in text
- Coreference Resolution (CO)
  - Identifies identity relations between entities in texts
- Template Element construction (TE)
  - Adds descriptive information to NE results
- Scenario Template production (ST)
  - Instantiate scenarios using TEs





## Examples

- NE:
  - Thenia (Location), Algiers (Location), May 22 (Date), Wednesday (Date), etc.
- CO:
  - a powerful earthquake and the quake
  - ALGIERS and Algiers
- TE:
  - entity descriptions: Thenia is in Algeria, Algiers is capital of Algeria
- ST
  - combine entities in one scenario (as shown in the example)





# Question Answering (Hirschman&Gaizauskas'01)

- Given a document collection (can be the Web) and a natural language question
- Extract the answer from the document collection
  - answer can in principle be any expression but in many cases questions ask for specific types of information such as person names, location, dates, etc.
- Open domain in general
- Text Retrieval Conferences Question Answering Track responsible for advances in the field of system development and evaluation (since 1999)
- From 2008 the Text Analysis Conference





# QA Task (Voorhees'99)

- In the Text Retrieval Conferences (TREC)
   Question Answering evaluation, 3 types of questions are identified
- Factoid questions such as:
  - "Who is Tom Cruise married to?"
- <u>List</u> questions such as:
  - "What countries have atomic bombs?"
- Definition questions such as:
  - "Who is Aaron Copland?" or "What is aspirin?"
     (Changed name to "other" question type)





### Text Summarization (Mani'01)

- Given a document or set of documents
- Extract the most important content from it and present a condensed version of it
  - extracts vs abstracts
- Useful for decision making: read or not read; saves time; can be used to create surrogates; etc.
- Open domain, however domain knowledge proves important (e.g., scientific domain)
- Document Understanding Conferences (since 2000) contributed with much development in the field
- From 2008 the Text Analysis Conference





# Integration of technologies for background gathering (Gaizauskas&al07)

- Cubreporter Project
  - IR, QA, TS, IE
- Background gathering: the task of collecting information from the news wire and other archives to contextualise and support a breaking news story
- Backgrounder components
  - similar events in the past; role players' profiles; factual information on the event
- Collaboration with Press Association
  - 11 year archive with more than 8 million stories





# **Background Examples**

#### Breaking News

"Powerful earthquake shook Turkey today"

#### Past Similar Events

"Last year an earthquake measuring 6.3-magnitude hit southern Turkey killing 144 people."

#### Extremes

"Europe's biggest quake hit Lisbon, Portugal, on November 1, 1755, when 60,000 people died as the city was devastated and giant waves 10 metres high swept through the harbour and on to the shore."

#### Definitions

"Quakes occur when the Earth's crust fractures, a process that <u>can be</u> <u>caused</u> by volcanic activity, landslides or subterranean collapse. The resulting plates grind together causing the tremors.





#### **Text Analysis Resources**

- General Architecture for Text Engineering
  - (http://gate.ac.uk)
  - Tokenisation, Sentence Identification, POS tagging, NE recognition, etc.
- SUPPLE Parser
  - (http://nlp.shef.ac.uk/research/supple)
  - syntactic parsing and creation of logical forms
- Summarization Toolkit
  - (http://www.dcs.shef.ac.uk/~saggion)
  - Single and multi document summarization
- Lucene
  - (http://lucene.apache.org)
  - Text indexing and retrieval





# **Summarization System**

- Scores sentences based on numeric features in both single and multi-document cases
  - position of sentence, similarity to headline, similarity to cluster centroid, etc.
  - values are combined to obtain the sentence score
  - single-document summaries and summaries for "related" stories
- Press Association profiles are automatically identified
- Other profiles created using QA/summarization techniques





# **Question Answering**

- Passage (i.e., paragraph) retrieval using question
- Question and passage analysis using a parser (SUPPLE)
  - semantic representation
  - identification of expected answer type (EAT)
  - each "entity" in a sentence is considered a <u>candidate</u> answer
- Answer candidates in passages scored using
  - sentence score (overlap with question)
  - "similarity" of <u>candidate</u> answer to EAT
  - count relations between candidate and "question entities"
  - merge scores across passages and select candidate with highest score





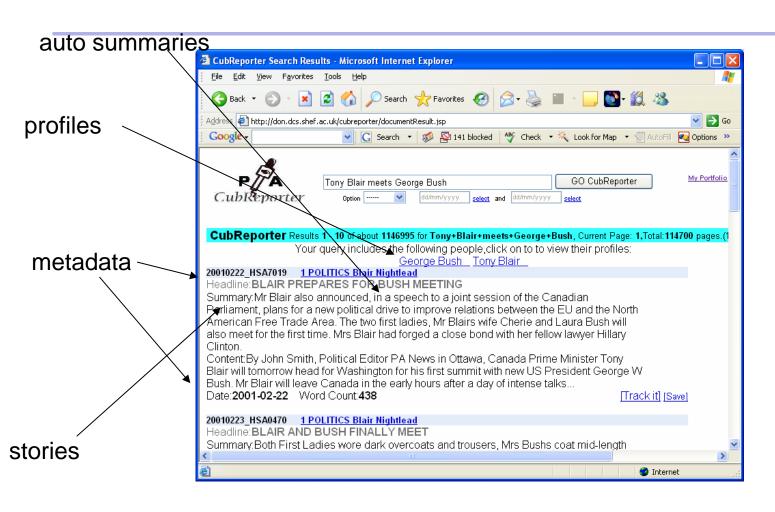
# Semantic Representations

- To search for "similar events"
- Leading paragraphs are parsed using SUPPLE and semantic representations created
  - "The head of Australia's biggest bank resigned today"
  - head(e2), name(e4,'Australia'), country(e4), of(e3,e4), bank(e3), adj(e3,biggest), of(e2,e3), resign(e1), lsubj(e1,e2)
- Database records are created and used to support similar event search
- We can search for "resignation" events
  - "resign" -> leave job = quit, renounce, leave office
  - "head" -> person in charge = chief





# Finding Stories

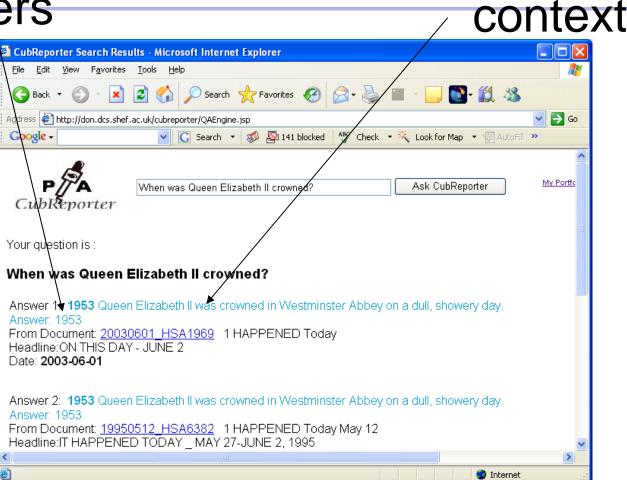






# **Getting Answers**

answers

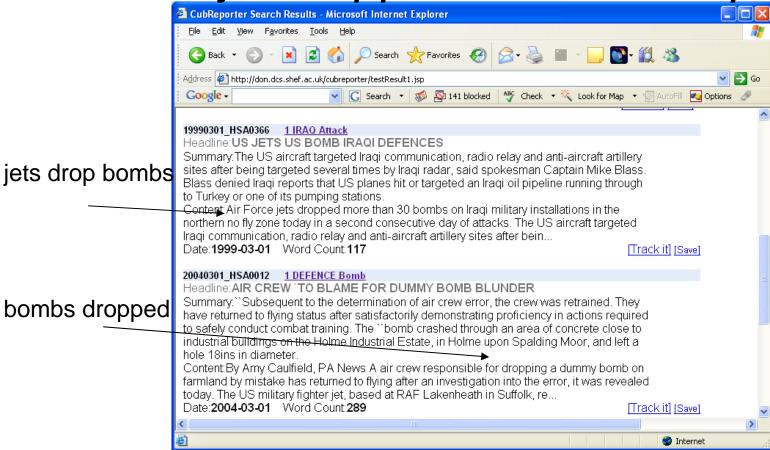






# **Getting Similar Events**

"jet dropped bomb in Iraq"







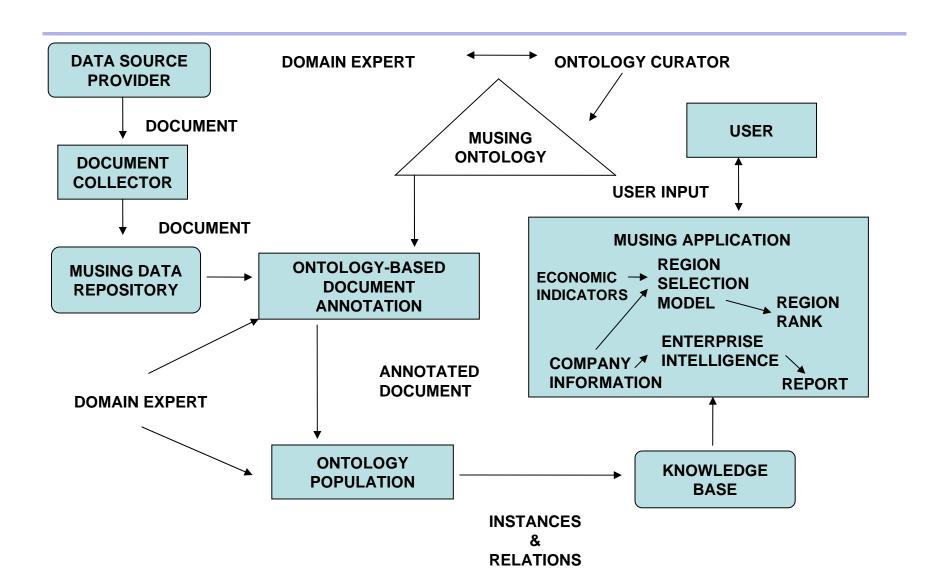
# Extracting information for business intelligence applications

- MUSING Project 6FP European Commission (ICT)
  - integration of natural language processing and ontologies for business intelligence applications
  - extraction of <u>company information</u>
  - extraction of <u>country/region information</u>
  - identification of <u>opinions</u> in text for company reputation





## Ontology-based IE in MUSING







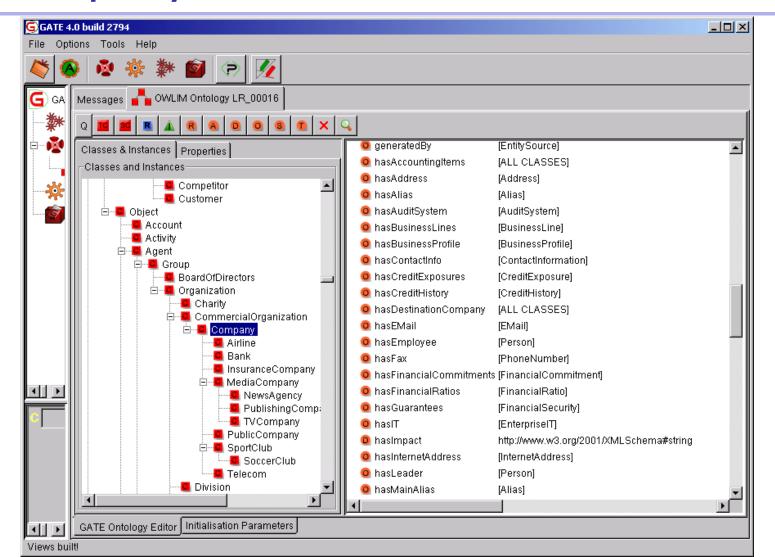
#### Data Sources in MUSING

- Data sources include balance sheets, company profiles, press data, web data, etc. (some private data)
  - News papers from Italian financial news provider
  - Companies' web pages (main, "about us", "contact us", etc.)
  - Wikipedia, CIA Fact Book, etc.
- Ontology is manually developed through interaction with domain experts and ontology curators
  - It extends the PROTON ontology and covers the financial, international, and IT operational risk domain





### Company Information in MUSING

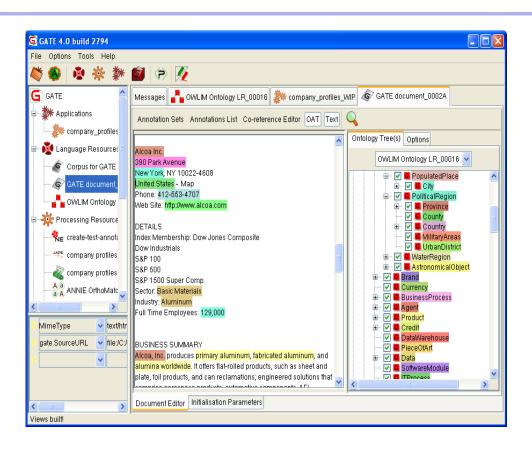






### **Extracting Company Information**

- Extracting information about a company requires for example identify the Company Name; Company Address; Parent Organization; Shareholders; etc.
- These associated pieces of information should be asserted as properties values of the company instance
- Statements for populating the ontology need to be created ( "Alcoa Inc" hasAlias "Alcoa"; "Alcoa Inc" hasWebPage "http://www.alcoa.com", etc.)







# General Architecture for Text Engineering – GATE (Cunningham&al'02)

- Framework for development and deployment of natural language processing applications
  - http://gate.ac.uk
- A graphical user interface allows users (computational linguists) access, composition and visualisation of different components and experimentation
- A Java library (gate.jar) for programmers to implement and pack applications





# Component Model

- Language Resources (LR)
  - data
- Processing Resources (PR)
  - algorithms
- Visualisation Resources (VR)
  - graphical user interfaces (GUI)
- Components are extendable and user-customisable
  - for example adaptation of an information extraction application to a new domain
  - to a new language where the change involves adaptation of a module for word recognition and sentence recognition





#### **Documents in GATE**

- A document is created from a file located somewhere in your disk or in a remote place or from a string
- A GATE document contains the "text" of your file and sets of <u>annotations</u>
- When the document is created and if a format analyser for your type is available "parsing" (format) will be applied and annotations will be created
  - xml, sgml, html, etc.
- Documents also store features, useful for representing metadata about the document
  - some features are created by GATE
- GATE documents and annotations are LRs





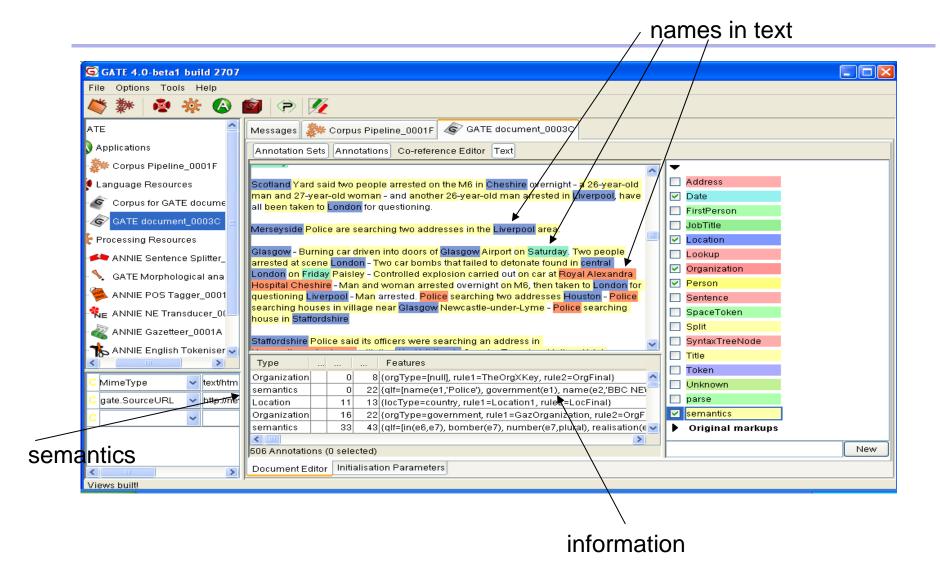
#### **Documents in GATE**

- Annotations have
  - types (e.g. Token)
  - belong to particular annotation sets
  - start and end offsets where in the document
  - features and values which are used to store orthographic, grammatical, semantic information, etc.
- Documents can be grouped in a <u>Corpus</u>
- Corpus is other language resource in GATE which implements a set of documents





#### Documents in GATE







#### **Annotation Guidelines**

- People need clear definition of what to annotate in the documents, with examples
- Typically written as a guidelines document
- Piloted first with few annotators, improved, then "real" annotation starts, when all annotators are trained
- Annotation tools require the definition of a formal DTD (e.g. XML schema)
  - What annotation types are allowed
  - What are their attributes/features and their values
  - Optional vs obligatory; default values





#### **Annotation Schemas**

```
<?xml version="1.0"?>
<schema
   xmlns="http://www.w3.org/2000/10/XMLSchema">
   <!-- XSchema definition for email--->
   <element name="Email" />
</schema>
```





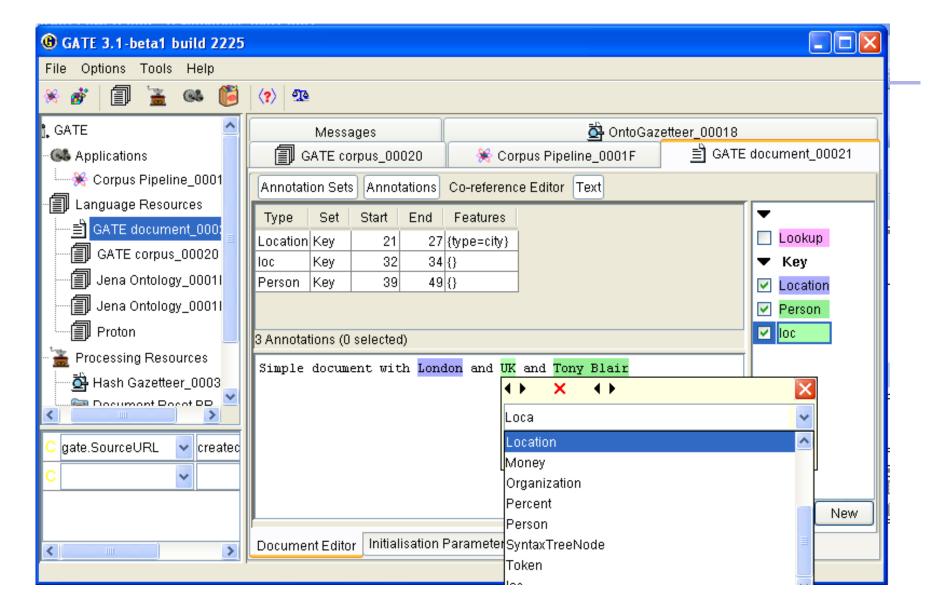
### **Annotation Schemas**

```
<?xml version="1.0"?>
<schema xmlns="http://www.w3.org/2000/10/XMLSchema">
   <!-- XSchema definition for token-->
   <element name="Address">
    <complexType>
     <attribute name="kind" use="optional">
     <simpleType>
      <restriction base="string">
         <enumeration value="email"/>
         <enumeration value="url"/>
         <enumeration value="phone"/>
         <enumeration value="ip"/>
         <enumeration value="street"/>
         <enumeration value="postcode"/>
         <enumeration value="country"/>
         <enumeration value="complete"/>
             </restriction> ...
```



#### Manual Annotation in GATE GUI

RuSSIR







#### Annotation in GATE GUI

# The following tasks can be carried out manually in the GATE GUI:

- Adding annotation sets
- Adding annotations
- Resizing them (changing boundaries)
- Deleting
- Changing highlighting colour
- Setting features and their values





# Preserving and exporting results

- Annotations can be stored as stand-off markup or in-line annotations
- The default method is standoff markup, where the annotations are stored separately from the text, so that the original text is not modified
- A corpus can also be saved as a regular or searchable (indexed) datastore





#### Corpora and System Development

- "Gold standard" data created by manual annotation
- Corpora are divided typically into a training, sometimes testing, and unseen evaluation portion
- Rules and/or ML algorithms developed on the training part
- Tuned on the testing portion in order to optimise
  - Rule priorities, rules effectiveness, etc.
  - Parameters of the learning algorithm and the features used
- Evaluation set the best system configuration is run on this data and the system performance is obtained
- No further tuning once evaluation set is used!





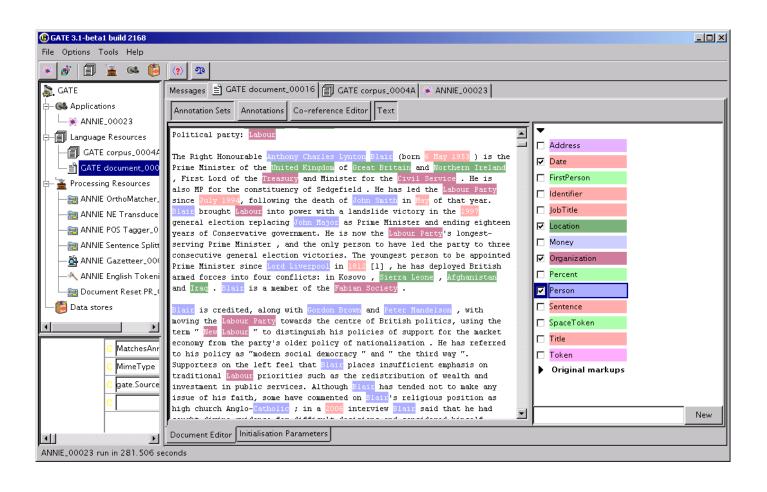
#### **Applications in GATE**

- Applications are created by sequencing processing resources
- Applications can be run over a Corpus of documents – <u>corpus pipeline</u>
  - so each component is applied to each document in the corpus in sequence
- Applications may not have a corpus as input, but different parameters – <u>pipeline</u>





# Name Entity Recognition







## **Text Processing Tools**

- Document Structure Analysis
  - different document parsers take care of the structure of your document (xml, html, etc.)
- Tokenisation
- Sentence Identification
- Parts of speech tagging
- Morphological analysis
- All these resources have as runtime parameter a GATE document, and they will produce annotations over it
- Most resources have initialisation parameters





#### Creole

- a Collection of REusable Objects for Language Engineering ~ Language Resources + Processing Resources
- creole.xml provides details about available components, the java class that implements the resource, and jar file where it is found



</RESOURCE>



#### Example of resource

```
<RESOURCE>
                                   <NAME>ANNIE English Tokeniser</NAME>
                                   <CLASS>gate.creole.tokeniser.DefaultTokeniser</CLASS>
                                   <COMMENT>
                                                     A customisable English tokeniser
                                                     (http://gate.ac.uk/sale/tao/#sec:en-tokeniser).
                                   </COMMENT>
                                   <PARAMETER NAME="document"
                                                     COMMENT="The document to be tokenised" RUNTIME="true">
                                                     gate.Document
                                   </PARAMETER>
                                   <PARAMETER NAME="annotationSetName" RUNTIME="true"
                                                     COMMENT="The annotation set to be used for the generated annotations"
                                                     OPTIONAL="true">
                                                     java.lang.String
                                   </PARAMETER>
                                   <PARAMETER NAME="tokeniserRulesURL"
                                                     DEFAULT="resources/tokeniser/DefaultTokeniser.rules"
                                                     COMMENT="The URL for the rules file" SUFFIXES="rules">
                                                     java.net.URL
                                   </PARAMETER>
                                   <PARAMETER NAME="transducerGrammarURL"
                                                     DEFAULT="resources/tokeniser/postprocess.jape"
                                                     COMMENT="The URL for the postprocessing transducer"
                                                     SUFFIXES="jape">
                                                     java.net.URL
                                   </PARAMETER>
                                   <PARAMETER NAME="encoding"
                                                     COMMENT="The encoding used for reading the definitions"
                                                     DEFAULT="UTF-8">
                                                     java.lang.String
                                   </PARAMETER>
                                   <ICON>tokeniser</ICON>
```





#### **Tokenisation**

- Identify different words in text: numbers, symbols, words, etc.
  - not only sequences between spaces or separators (in English)
  - 2,000 is not "2" "," "000"
  - I've is "I" and "'ve"
  - \$20 is "\$" and "20"





#### Tokenisation in GATE

- Rule-based LHS > RHS
  - LHS is a regular expression over character classes
  - RHS specifies annotation to be created and features to be asserted for the created annotation
    - "DECIMAL\_DIGIT\_NUMBER"+ >Token;kind=number;
    - (SPACE\_SEPARATOR) > SpaceToken; kind=space;
    - (CONTROL) >SpaceToken;kind=control;
    - "UPPERCASE\_LETTER" (LOWERCASE\_LETTER
       (LOWERCASE\_LETTER|DASH\_PUNCTUATION|FORMAT)\*
       )\* > Token;orth=upperInitial;kind=word;
- Tokeniser produces a <u>Token</u> type of annotation





#### Tokenisation in GATE

- Features produced by the tokeniser
  - string: the actual string of the token
  - orth: orthographic information
  - length: the length of the string
  - kind: the type of token (word, symbol, punctuation, number)
- SpaceToken is another annotation produced
  - features: kind (control or space), length, string





## Sentence Splitter

end of sentence

not end of sentence

- Decide where a sentence/ends
  - "The court ruled that Dr. Smith was innocent."
- A rule based mechanism
  - uses a list of known abbreviations to help identify end of sentence uses (e.g. Dr)
  - a period is a sentence break if it is preceded by a non-abbreviation and followed by an uppercase common word
- The Splitter in GATE produces a <u>Sentence</u> type of annotation





#### Parts of Speech Tagging (Hepple'00)

- Associate a part of speech tag to each word
  - tags from the Penn Treebank including punctuation
- Based on Brill's tagger but a different learning approach used for rule acquisition – no need for re-annotating the corpus at each iteration during learning
- Two steps during tagging
  - initial guess based on lexicon (contain most likely tag)
  - correction based on a list of rules (contextual)





# POS tags used

```
CC - coordinating conjunction: "and", "but", "nor", "or", "yet", plus, minus, less, times (multiplication), over (division). Also "for" (because) and "so" (i.e., "so
that").
CD - cardinal number
DT - determiner: Articles including "a", "an", "every", "no", "the", "another", "any", "some", "those".

EX - existential there: Unstressed "there" that triggers inversion of the inflected verb and the logical subject; "There was a party in progress".
FW - foreign word
IN - preposition or subordinating conjunction
JJ - adjective: Hyphenated compounds that are used as modifiers; happy-go-lucky.

JJR - adjective - comparative: Adjectives with the comparative ending "-er" and a comparative meaning. Sometimes "more" and "less".

JJS - adjective - superlative: Adjectives with the superlative ending "-est" (and "worst"). Sometimes "most"and "least".
JJSS - -unknown-, but probably a variant of JJS
-LRB- - -unknown-
LS - list item marker: Numbers and letters used as identifiers of items in a list.
MD - modal: All verbs that don't take an "-s" ending in the third person singular present: "can", "could", "dare", "may", "might", "must", "ought", "shall", "should",
"will", "would".
NN - noun - singular or mass
NNP - proper noun - singular: All words in names usually are capitalized but titles might not be.
NNPS - proper noun - plural: All words in names usually are capitalized but titles might not be.
NNS - noun - plural
PDT - predeterminer: Determinerlike elements preceding an article or possessive pronoun; "all/PDT his marbles", "quite/PDT a mess".
POS - possesive ending: Nouns ending in "s" or """.
PP - personal pronoun
PRPR$ - unknown-, but probably possessive pronoun
PRP - unknown-, but probably possessive pronoun
PRP - unknown-, but probably possessive pronoun
PRP$ - unknown, but probably possessive pronoun, such as "my", "your", "his", "his", "its", "one's", "our", and "their".
RB - adverb: most words ending in "-ly". Also "quite", "too", "very", "enough", "indeed", "not", "-n't", and "never".
RBR - adverb - comparative: adverbs ending with "-er" with a comparative meaning.
RBS - adverb - superlative
RP - particle: Mostly monosyllabic words that also double as directional adverbs.
STAART - start state marker (used internally)
SYM - symbol: technical symbols or expressions that aren't English words.
TO - literal to
UH - interjection: Such as "my", "oh", "please", "uh", "well", "yes".

VBD - verb - past tense: includes conditional form of the verb "to be"; "If I were/VBD rich...".
VBG - verb - gerund or present participle
VBN - verb - past participle
VBP - verb - non-3rd person singular present
VB - verb - base form: subsumes imperatives, infinitives and subjunctives.
VBZ - verb - 3rd person singular present
WDT - wh-determiner
WP$ - possesive wh-pronoun: includes "whose"
WP - wh-pronoun: includes "what", "who", and "whom".
WRB - wh-adverb: includes "how", "where", "why". Includes "when" when used in a temporal sense.
```





# Parts of Speech Tagging

- Two resources
  - lexicon collected from corpus with <word, list of valid tags>
    - employs VBZ
    - empty JJ VB VBP
  - some heuristics for unknown words
  - rules for correcting tagging mistakes
    - NN VBG PREVWD before
  - rules instantiate patterns such as:
    - Change tag A to tag B if Condition
- The GATE tagger produces a feature <u>category</u> for each token in the document, the value of the feature is the name of the POS tag





#### Morphological Analysis in GATE

- For each noun and verb in the document identifies lemma and affix which are stored in the Token annotation ('root', 'affix')
- A set of rules for regular cases is used
- A set of irregular cases which explicitly indicate how to decompose the word is also used





#### Stemming in GATE

- Removing prefixes and suffixed of a word
  - produces a feature 'stem' in the Token annotation
  - John -> stem=john
  - tells -> stem=tell
  - considered -> stem=consid (root=consider)
  - leaving -> stem=leav (root=leave)
  - had -> stem=had (root=have)
- Available for English and other languages (e.g. Spanish)





## Named Entity Recognition

- It is the cornerstone of many NLP applications in particular of IE
- Identification of named entities in text
- Classification of the found strings in categories or types
- General types are Person Names, Organizations, Locations
- Others are Dates, Numbers, e-mails, Addresses, etc.
- Domains may have specific NEs: film names, drug names, programming languages, names of proteins, etc.





#### NER problems



- There are problems even with well known categories
- "Ambrose Chapel" it's not a <u>name</u> it is a <u>place!!!!</u>
- Ambiguity is one problem
  - "Paris" can be a city or a person
    - Paris (for Paris Hilton, the Person); Paris Hilton hotel (the place)
  - London can be a place or an organization (the government)





#### Approaches to NER

- Two approaches: (1) Knowledge-based based on humans defining rules; (2) Machine learning approach, possibly using an annotated corpus
- Knowledge-based approach
  - Word level information is useful in recognising entities:
    - capitalization, type of word (number, symbol)
  - Specialized lexicons (Gazetteer lists) usually created by hand; although methods exist to compile them from corpora
    - List of known continents, countries, cities, person first names
    - On-line resources are available to pull out that information





# Approaches to NER

- Knowledge-based approach
  - rules are used to combine different evidences
  - a known first name followed by a sequence of words with upper initial may indicate a person name
  - a upper initial word followed by a company designator (e.g., Co., Ltd.) may indicate a company name
  - a cascade approach is generally used where some basic names are first identified and are latter combined into more complex names





## Approaches to NER

- In GATE Gazetteers lists entries may contain some useful semantic information
  - for example one may associate some features and values to entry names
  - features can be used in grammars or can be used to enrich system output
  - gazetteer lists are organized in index files





#### Gazetteers in GATE

- Lists store keywords (one keyword per line)
  - list of male names (person\_male.lst)

Aaron

**Abraham** 

....

- Set of lists compiled and a finite state machine is created which operates on the strings
- The machine produces annotations of type <u>Lookup</u> when the keyword is found in text
- 60k entries in 80 types:
  - organization; artifact; location; amount\_unit; manufacturer;





#### Gazetteer in GATE

- Sets of lists are organized in a main lists file
- Each list specifies attributes <u>majorType</u> and <u>minorType</u> and <u>language</u>, having major and minor types gives some flexibility to grammar rules

```
government.lst:organization:government
```

department.lst:organization:department

person\_male.lst:person\_first:male

person\_female.lst:person\_first:female

(look into gate/plugins/ANNIE/gazetteers for examples)

- Attributes are used to help identification of more complex entities (for example discriminating when possible between a male or female name)
- List entries may be entities or parts of entities, or they may contain contextual information (e.g. job titles often indicate people)





#### Named Entity Grammar in GATE

- Implemented in the JAPE language (part of GATE)
  - Regular expressions <u>over annotations</u>
  - Provide access and manipulation of annotations produced by other modules
- Rules are stored in grammar files
- Grammar files are compiled into Finite State Machines
- A main grammar files specifies how different grammars should be executed (phases)
  - constitute a cascade of FSTs over annotations





#### NER in GATE

- Rules are <u>hand-coded</u>, so some linguistic expertise is needed here
- uses annotations from tokeniser, POS tagger, and gazetteer modules
- use of contextual information
- rule priority based on pattern length, rule status and rule ordering
- Common entities: persons, locations, organisations, dates, addresses.





#### JAPE Language

- A JAPE grammar rule consists of a left hand side (LHS) and a right hand side (RHS)
  - LHS= what to match (the pattern)
  - RHS = how to annotate the found sequence
  - LHS - > RHS
- A JAPE grammar is a sequence of grammar rules
- Grammars are compiled into finite state machines
- Rules have priority (number)
- There is a way to control how to match
  - options parameter in the grammar files





#### LHS of JAPE rules

- The LHS of the rule contains patterns to be matched, in the form of annotations (and optionally their attributes).
- Annotation types to be recognized must be declared at the beginning of the phase
- Annotations may be combined using traditional operators [ | \* + ?]





# Referring to annotation in JAPE

- Token; Token.string == "..."; Lookup; Lookup.majorType == "..."; Person; etc.
- {Token.kind == word, Token.length == 2}
- ({Token.kind == word} | {Token.length == 2})
- {Token.kind == word} {Token.kind == word}
- ({Token.orth == upperInitial})+
  {Lookup.majorType == location}





#### LHS of JAPE rules

- There is no negative operator
- More than one pattern can be matched in a single rule
- Left and right context (not to be annotated)
   can be matched
- LHS has labels to be referred to in RHS





#### **Examples of LHS patterns**

- //identify a token with upper initial
  ({Token.orth == upperInitial}):upper
- //recognise a sequence of one upper initial word followed by a location designator (e.g. Ennerdale Lake)
   ({Token.orth == upperInitial}
   {Lookup.majorType == loc\_designator}) :location
- //same but with upper initial or all capitals
  (({Token.orth == upperInitial}|{Token.orth == allCaps})
  {Lookup.majorType == loc\_designator}) :location





# Example of RHS

```
({Token.orth == upperInitial}
      {Lookup.majorType == lake_designator}) :location
→
:location.Location = { type = "lake" }
```

 Indicates annotation type to be produced 'Location' and features and values for that annotation type





#### Macros in JAPE grammars

```
Macro: ONE DIGIT
(\{Token.kind == number, Token.length == "1"\})
Macro: TWO DIGIT
(\{Token.kind == number, Token.length == "2"\})
Macro: FOUR DIGIT
({Token.kind == number, Token.length == "4"})
Macro: DAY MONTH NUM
(ONE DIGIT | TWO DIGIT)
In the LHS of the rule one can use the macro name:
   (DAY_MONTH_NUM):annotate -> :annotate.DAY = {}
```





# Example of RHS (context)

Rule: Date

```
({Token.string == "Date"} {Token.string == ":"}):context
(({Token.kind == "number", Token.length == "2"})
 ({Token})
({Token.kind == "number", Token.length == "2"})
 ({Token})
({Token.kind == "number", Token.length == "2"})):annotate
-->
:annotate.Date = { type = "dd/mm/yy format" }
```





#### JAPE Grammar

 In a file with name something.jape we write a Jape grammar (phase)

```
Phase: example1
Input: Token Lookup
Options: control = appelt
Rule: PersonMale
Priority: 10
{Lookup.majorType == first_name, Lookup.minorType == male}
({Token.orth == upperInitial})*
):annotate
-->
:annotate.Person = { gender = male }
....(more rules here)
```





# Main JAPE grammar

 Combines a number of single JAPE files in general named "main.jape"

MultiPhase: CascadeOfGrammars

Phases:

grammar1

grammar2

grammar3





# Further processing in RHS

- Java code can be included in the RHS of the rule
- It is a powerful mechanism which can help add semantic information to the annotations
  - for example extracting information from the context





# Available Java objects

- bindings: labels used in LHS are available
- doc: the GATE document which is being process
- annotations: all GATE document annotations produced until that stage
- inputAS, outputAS: phase input and output annotations





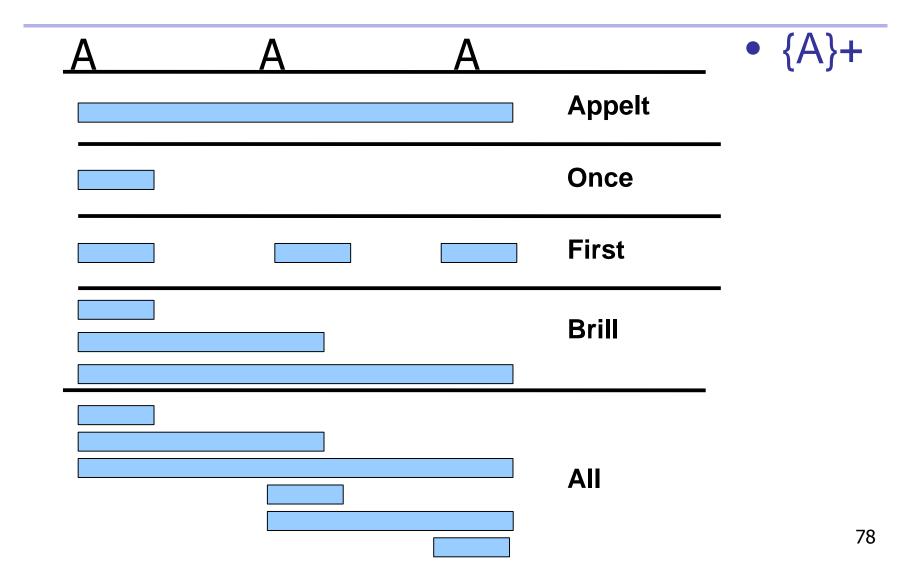
#### JAPE Application modes

- Matching control for rules
  - Brill (fires all matches)
  - First (shortest match fires)
  - Once (Phase exits after first match)
  - All (as for Brill, but matching continues from offset following the current one, not from the end of the last match)
  - Appelt (priority ordering: longest match fires, then explicit rule priority, then first defined rule fires)





### **JAPE Application Modes**







### Using phases

- Grammars usually consist of several phases, run sequentially
- A definition phase (conventionally called main.jape) lists the phases to be used, in order
- Only the definition phase needs to be loaded
- Temporary annotations may be created in early phases and used as input for later phases
- Annotations from earlier phases may need to be combined or modified





#### Coreference Resolution

#### Name coreference

- matches similar names in text, e.g. "Dr. Jacob Smith" and "Smith"
- creates a 'matches' annotation which allows you to extract a chain of 'equivalent' names

#### Pronominal coreference

 solves references to named entities of pronouns in English (tokens marked with POS category PRP or PRP\$)





#### Coreference Resolution

- Orthographic co-reference can improve NE results by assigning entity type to previously unclassified names, based on relations with classified NEs
- May not reclassify already classified entities
- Classification of unknown entities very useful for surnames which match a full name, or abbreviations, e.g. "Bonfield" will match "Sir Peter Bonfield"





# **ANNIE System**

- A Nearly New Information Extraction System
  - recognizes named entities in text
  - "packed" application combining/sequencing the following components: document reset, tokeniser, splitter, tagger, gazetteer lookup, NE grammars, name coreference
  - can be used as starting point to develop a new named recogniser





#### Performance Evaluation

- Evaluation metric mathematically defines how to measure the system's performance against a human-annotated, gold standard
- Scoring program implements the metric and provides performance measures
  - For each document and over the entire corpus
  - For each type of NE





#### The Evaluation Metric

- Precision = correct answers/answers produced
- Recall = correct answers/total possible correct answers
- Trade-off between precision and recall
- F-Measure =  $(\beta^2 + 1)PR / \beta^2R + P$ [van Rijsbergen 75]
- ullet eta reflects the weighting between precision and recall, typically eta=1





#### **Document Indexing**

- Indexing with Lucene
  - Populate a corpus
  - Create a Data Store (java serialisation)
  - Save corpus to DS
    - the corpus become "indexable" by Lucene
  - Index the corpus using document content
  - To search use the Information Retrieval plug-in
    - SearchPR processing resource (put it in a "pipeline")
    - Specify parameters of search (corpus, query, etc.) and run
    - double clicking on SearchPR displays the results





#### Annotations in Context (ANNIC)

- Create a linguistic/semantic index
  - Create a Lucene Searchable DS
  - Populate a corpus
  - Apply ANNIE to the corpus
  - Save corpus to DS
  - Search in Context in the DS GUI





# Machine Learning Approach

- Given a corpus annotated with named entities we want to create a classifier which decides if a string of text is a NE or not
  - ...<person>Mr. John Smith</person>...
  - ...<date>16th May 2005</date>
- The problem of recognising NEs can be seen as a classification problem





### Machine Learning Approach

- Each named entity instance is transformed for the learning problem
  - ...<person>Mr. John Smith</person>...
  - Mr. is the beginning of the NE person
  - Smith is the end of the NE person
- The problem is transformed in a binary classification problem
  - is token begin of NE person?
  - is token end of NE person?
- Context is used as features for the classifier





# Parsing with SUPPLE (Gaizauskas&al'05)

- Sheffield University Prolog Parser for Language Engineering
- A bottom-up parser for English which produces syntactic and semantic sentence respresentations
- An attribute-value context-free grammar of English is used to derive syntactic representations (it includes a question grammar for QA applications)
  - categories in the grammar have attributes and values which can be instantiated during parsing





- The grammar covers the following constituents
  - prepositional phrases; noun phrases; core verbs; verb phrases; relative clauses; sentences; questions
- The input to the parsing process is a 'chart' where both lexical items and multiword expressions (named entities) are allowed
- The output is the best possible 'parse' of the sentence, this can be partial





- Semantics is constructed compositionally as the sentence is parsed
- nouns and verbs are represented as normalised unary predicates (cat, eat, etc.)
- Identifiers (ei) are used to refer to an entity or an event and are produced for each noun and verb
  - cat(e1), eat(e2)
- binary predicates represent relations or attribute values of the entities or events; they are a fixed inventory used to represent grammatical and semantic relations
  - Isubj(X,Y), lobj(X,Z), of(X,Y), name(X,Z),...





#### Example

- Tony Blair meets U.S. President Bush.
  - identifies Tony Blair and Bush as Person type and U.S. is a Location type
  - wraps those constituents so that SUPPLE does not have to analyse them
  - rest of elements in the sentence are passed as words with POS, roots, number, gender, etc.

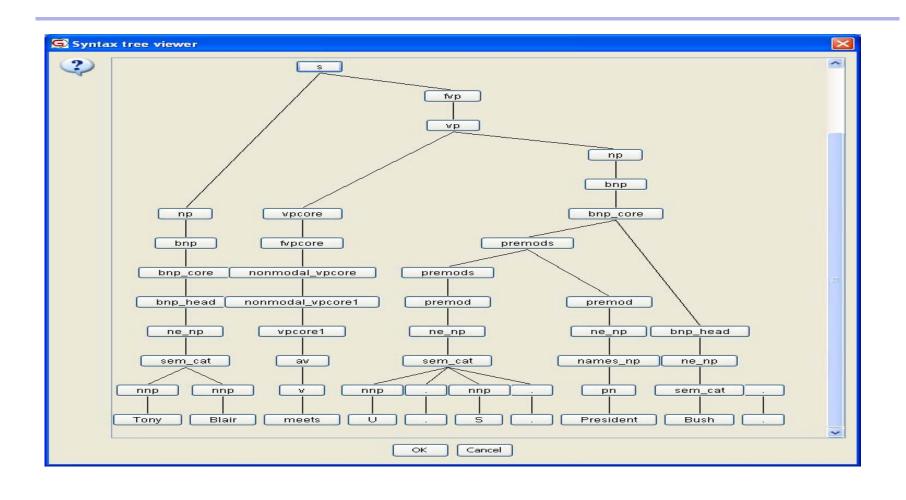




Syntactic Annotation (string) best\_parse=( s ( np ( bnp ( bnp\_core (
bnp\_head ( ne\_np ( sem\_cat "Tony Blair" ) ) )))))(fvp(vp(vpcore(fvpcore)( nonmodal\_vpcore ( nonmodal\_vpcore1 ( vpcore1 ( av ( v "meets" ) ) ) ) ) ) ) ( np ( bnp (bnp\_core (premods (premods ( premod ( ne\_np ( sem\_cat "U.S." ) premod (ne\_np (names\_np (pn "President" ))(bnp\_head(ne\_np(`sem cat "Bush"











Semantic Annotation (array of strings) qlf=[name(e2,'Tony Blair'), person(e2), realisation(e2,offsets(0,10)), meet(e1), time(e1,prèsent), aspect(é1,simple), voice(e1,active), lobj(e1,e3), name(e3,'Bush'), person(e3), name(e4,'U.S.'), location(e4), country (e4), realisation (e4, offsets (17, 21)), qual(e3,e4), ne\_tag(e5,offsets(22,31)), name(e5,'President'), realisation(e5,offsets(22,31)), qual(e3,e5), realisation(e3,offsets(17,36)), realisation(e1,offsets(11,36)), lsubj(e1,e2)]





- A wrapper is provided in GATE
  - given a text which has been POS-tagged and Morphologically analysed, maps the tokens in each sentence to the input expected by SUPPLE
  - read the syntactic and semantic information from files and stores the information into the GATE documents as
    - parse, semantics, syntax tree nodes
- Can be run with SICStus prolog, SWI prolog, and PrologCafe (Java implementation)





### Summary of first part

- Examples of Information Access
   Applications: Cubreporter & Musing
- General Architecture for Text Engineering (GATE)
  - Components: LR & PR
  - Demonstration: GUI and Java programs
  - Applications for text processing and named entity recognition