

# A Framework for Text Mining Services

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## Abstract

The growth of online scientific literature, coupled with the growing maturity of text processing technology, has boosted the importance of text mining as a potentially crucial tool. However, there are several challenges to be addressed before sophisticated text mining services can be deployed within emerging workflow environments. Our work contributes at two levels. At the invocation level, we have developed a flexible XML-based pipeline architecture which allows non-XML processors to be readily integrated. At the description/discovery level, we have developed a broker for service composition, and an accompanying domain ontology, that leverage the OWL-S approach to service profiles.

## 1 Introduction

There is increasing interest in deploying text mining tools to assist researchers in various tasks. For example, a social historian may wish to query a gazetteer with location names retrieved from a collection of transcribed interviews in order to obtain parish codes; these can then be used in analysing aggregate regional statistics for the period. Another example is the astronomer who detects an X-ray signal in some region of the sky, and wants to investigate the online literature to see if any interesting infra-red signals were detected in the same region. A third example involves the scientist who has been given a list of 100 genes detected in a micro-array experiment, and wishes to trawl through the existing published research for papers which throw light on the function of these genes.

In such scenarios, we envisage the researcher using a workflow editor to incorporate one or more text mining services within a larger application. The service might be essentially a document classification tool which retrieves documents in response to a string of keywords. However, it might be more elaborate, such as an information extraction system which populates a database that is subsequently queried by some further service.

Typically, a text mining system that is deployed as a service will have been assembled out of modules, each of which carries out a specific natural lan-

guage processing (NLP) task. For example, an information extraction system might include modules for tokenization, part-of-speech tagging, named entity recognition, and relation detection. The *scientist end-user* will not care about the details of these modules, and we might assume therefore that the service is presented as a black box. Nevertheless, there are a number of reasons why we might care about the internal architecture. First, the external surface of a service might change according to the application: in one case, the output of the relation detector would be directed to a database service or a presentation layer; in a second case, the relation detector would be discarded, and the output of the named entity recognizer would be exposed as the external interface. For the sake of generality, therefore, we do not want to impose too rigid a grain-size on the components that can be offered as services.

Second, a *provider* of text mining services would want the ability to easily reconfigure systems for both testing and deployment purposes. In general, new application domains will require statistically based tools to be retrained, and rule-based tools to be provided with new rule sets; these will both potentially affect the overall system configuration. More importantly, ongoing rapid progress in the development of data-intensive text processing tools has the consequence that the top-performing systems will usually be heterogeneous, in the sense of incorpo-

rating processing modules developed by a variety of research groups. This argues strongly in favour of an underlying architecture which supports loosely-coupled, interchangeable processors. This in turn raises the possibility that a text mining system which is presented as a black box to the scientist end-user might in fact be decomposable for the computational linguist as a workflow of independent, reusable services—in other words, we might want to offer the computational linguist a workflow environment that allows her to rapidly prototype and test applications built from tools exposed as web services. In such a case, we could interoperate with the end user by providing a pre-packaged workflow that could be included within a larger application.

In order to explain our perspective on text mining services, we will start (Section 2) by describing our underlying XML-oriented approach to shallow text processing. In order to illustrate the kind of utility that our text processing tools support, we will focus on a task that arose in from collaborative work with EDINA<sup>1</sup> on the the JISC-funded **geoXwalk** project,<sup>2</sup> namely identifying placenames in text documents. We will then briefly describe how we have put together a series of text processing steps into a service-oriented workflow that contributes to the overall application. In Section 3, we turn to the question of how to use semantic characterizations of text processing services in order to help the user discover and compose the services into useful applications.

## 2 LT-XML Language Processing

### 2.1 Overview

The LT-XML [11] architecture for text processing was designed to support loosely coupled processors of the kind just alluded to, but using command line programs communicating via Unix pipes. In order to be pipelined in this way, all processors must share a common data format, and the choice of XML allows us to accommodate a variety of models of linguistic annotation. Following standard Unix philosophy, individual processors each carry out very specific tasks. However, the fact that they all stream their input and output in XML allows them to be freely combined, and different combinations of the same suite of processors can carry out a range of complex text processing tasks. Example applications developed over recent years include identifying gene and protein names in Medline abstracts [6, 7, 5], tokenizing technical terms in Medline ab-

stracts as a prelude to wide-coverage parsing [10, 9], extracting information about laptop computer specifications from web pages [12], identifying geographical locations in Scottish historical documents [19], and high-compression summarization of legal documents [8, 14].

To date, all NLP applications built within LT-XML have involved adding annotation to plain or HTML-formatted textual input. With relatively minimal assumptions about the appropriate data model for linguistic annotation, it is possible to adopt a simple processing model: each processor in a pipeline (except the very first) inspects the markup produced by the preceding processor, and modifies this or adds new markup of its own. Although the markup process need not be strictly monotonic,<sup>3</sup> in practice it almost always is. Figure 1 illustrates this point: in the topmost tree, the text has been annotated to show word (`<w>`) and sentence (`<s>`) units, while the bottom tree has additional annotation which marks part-of-speech information on words, and adds a named entity element (`<ne>`) of type *location*.

A base set of LT-XML text processors has been built using *lxtransduce*, a general purpose deterministic transducer which rewrites a stream of characters or XML elements according to a set of rules. The input is matched with regular expressions or XPath patterns. Character input can be arbitrarily modified and marked up with XML elements. XML input can be re-ordered and marked up with additional elements and attributes. Input can be matched against lexicons and the lexical information incorporated in the output. With suitable rule sets, *lxtransduce* allows us to construct tokenizers, named entity recognizers and partial parsers.

However, as argued earlier, a high priority is to allow heterogeneous systems, incorporating third-party—typically non-XML—processors into an XML-based pipeline. This requires us to use format converters (or ‘shims’ following [21]) to map out of, and back into, XML. Most of our legacy code uses *ad hoc* scripts *xmlperl*,<sup>4</sup> for this purpose, which gives us a powerful and flexible tool for rapid prototyping. However, we are also experimenting with XSLT stylesheets for the former, and with Yacc/Bison for the latter.

The creation of reliable shims is important since the performance of most text processing applications is usually limited by the performance of the weakest component. However, most NLP frameworks limit the use of processors available in an application to those that conform to their internal data interchange

<sup>1</sup><http://edina.ac.uk/>

<sup>2</sup><http://www.geoxwalk.ac.uk/>

<sup>3</sup>By ‘monotonic’, we mean that annotations can be added but not removed.

<sup>4</sup><http://www.cogsci.ed.ac.uk/~dmck/xmlstuff/>

format. Since state-of-the-art tools often use non-XML formats, shims must be created for those tools so that they can be integrated and thus obtain the optimal performance for the system as a whole.

However, when modularizing a text processor and creating its shim we do not want to limit its use to the XML data interchange format of a particular framework. However, by publishing the XML format used by the output and input of an processor (e.g., making its XML Schema or DTD publicly available), it should often be possible to create an XSLT stylesheet to translate between the XML data interchange formats used by the various frameworks and processors. If processors are wrapped using one of the popular XML data interchange formats, then only a few standard transformations from one common data interchange format to another may need to be created. By using the method of Schema inclusion, only the relevant parts of the XML data interchange formats need to be translated. This approach lets any NLP processor be used in any text processing workflow, provided that mappings can be created between the relevant parts of the data formats. For our particular pipeline we used the LT-XML data interchange format, and mappings can be created from that format to parts of other formats such as NITE.”

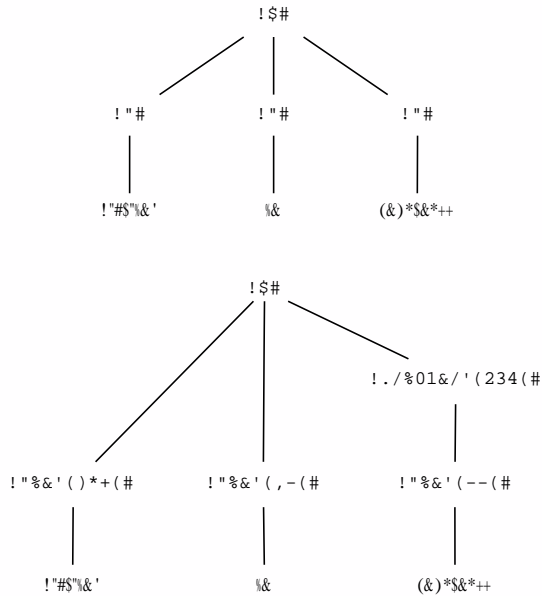


Figure 1: Incremental XML Annotation

## 2.2 GeoParsing

We will illustrate a typical usage of our tools by focussing on *geoparsing*—the task of automatically identifying location names in documents. This needs

to be seen within the broader context of working with spatial data.

A wide range of socio-economic data is regularly collected and published by statistical organizations and governments, such as statistics pertaining to population sizes, health, crime, or unemployment. Such datasets relate the primary variables to spatial information in various ways [17]; in particular, they use different *geographies*, ranging from explicit grid references or parish codes to implicit references such as postal addresses or association with a place-name. In order to support the process of relating data to a representation of its spatial scope (sometimes called *geo-referencing*), services such as *geoXwalk* [4] have been developed to translate between these different geographies.

Mapping across different systems of spatial reference allows for the fusion of different datasets, which in turn is a precondition for sophisticated mapping. For example, a social scientist studying the relationship between crime and deprivation in Scotland might want to draw a *thematic map* showing the strength of the spatial correlation between crimes occurring in a given place and standardized scores of deprivation in the same area [18]; the former might be based on address geo-referenced police incidence records and the latter on UK Census boundaries.

The particular system we describe here is intended to take HTML documents as input, and to output documents with the same HTML markup, but supplemented with `<span>` tags which visually highlight the placenames in the document—obviously, other presentation modes could be derived, given that the intermediate results are encoded in XML. Figure 2 gives a browser view of a document after location names have been marked up.

A recent experiment on using the system ( which incorporates an off-the-shelf maximum entropy tagger [3]) to recognize location names in the Statistical Accounts of Scotland has achieved a *f*-score (combined precision and recall) in recognition of over 94%; see [19] for more details.

In addition, the system uses the location names extracted from the source text to query an online gazetteer service.<sup>5</sup> For example, a placename query for *Sudbury* will retrieve an XML record which contains an element called *standard-reports* which is made up of a sequence of *gazetteer-standard-report* elements, one for each gazetteer entry. Each of these contains a bounding-box element which contains its bounding coordinates. There is also an element to describe the location sub-type: e.g., there are *Sudbury* entries variously annotated as ‘parish admin’,

<sup>5</sup>hosted by EDINA at <http://dearg.ucs.ed.ac.uk:9101/adlcrosswalk/>.

### Suffolk

There is something very English about Suffolk. Here you will find the quintessential English village. Along the river Stour south of Lavenham is 'Constable Country', home of England's most famous landscape painter and the subject of his glorious paintings which so poignantly captured rural English life. The Stour flows peacefully through a chalky valley with willow and poplar lining lush water-meadows, and walks between Sudbury and Flatford capture the essential 'Englishness' of the scenery. Flatford is, of course, the location for one of Constable's most famous paintings and is now a major tourist attraction. Upstream at Sudbury is the birthplace of another great English painter, Thomas Gainsborough. The long distance route The Stour Valley Path follows the river all the way from Newmarket to Manningtree.



Figure 2: Web page marked up with location name

'ward', and 'village or hamlet'. The module which sends queries to the gazetteer uses the query results to create a new XML document containing a list of virtual\_locations found in the input document: 'virtual' because the same name is often shared by more than one place; for example, there exist around 30 places called 'Aberdeen' on earth, and England alone has two 'Leeds'. We are currently working on methods to resolve these ambiguities automatically [16]. This output is illustrated in Figure 2.2.

```
<virtual_location>
<occurrence>
<text>Sudbury</text>
<index>49</index>
</occurrence>
<occurrence>
<text>Sudbury</text>
<index>65</index>
</occurrence>
<standard-reports>
.....
</standard-reports>
</virtual_location>
```

That is, a virtual\_location lists the tokens which have been marked as placenames in the text, together with the token identifiers, and then gives the standard-reports element from the output of the gazetteer query. The information in this XML file can be display in an interactive window which allows the user to disambiguate the intended referent of the placename, to select places of particular interest, ask for more information about them from other sources, and request a map which displays the selected places.

## 2.3 Service Architecture

The initial implementation of the geoparser described above ran as a single Unix script on a single system. However, by making its major modules available as web services, the entire system can become distributed across multipl platforms and components can be reused and interchanged.

Nevertheless, the foundation of many text processing components as pipelines has ramifications for their development as web services. Many web services are currently based on well-defined typed arguments, typically with the service itself built over a strongly-typed programming language such as Java. In contrast, text processing web services are almost always document-centered, exchanging and passing linguistic information as XML input and output documents embedded within SOAP messages. While these documents may be validated by a schema, the crucial information needed by the web service is embedded within a single XML input document itself, not as multiple strongly-typed parameters. Also, a single XML document is usually returned as the output. To facilitate this document-centric approach to web service workflows, we created our own web service wrappers and servers.

The workflow involves four major steps:

**Tokenization:** Identification of words and sentences in the input text.

**Location Tagging with a classifier:** Using the maximum entropy tagger to markup location names.

**Location Tagging by Lexicon:** Using a local lexicon of location names to markup additional locations not identifier by the tagger.

**Gazetteer Query:** Sending location names retrieved from the text to a gazetteer resource.

In addition, the workflow is initiated by a step that converts HTML markup into XML, and terminated by a step that re-converts the XML back into HTML.

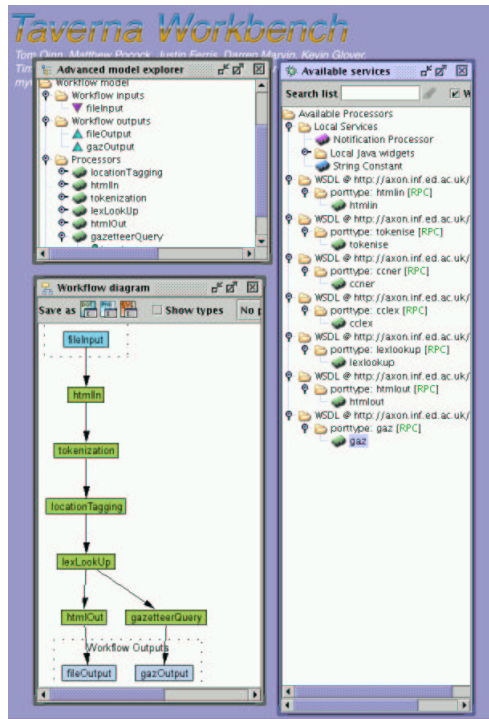


Figure 3: View of Location Tagging Workflow

Each of these steps is implemented as shell script on Linux, and consists of a pipeline of command-line programs. The scripts are then wrapped as web service using a utility called *lxsoap*. A simple configuration file acts as the input to an XSLT transformation that creates a WSDL file describing the service; it also allows *lxsoap* to map between SOAP messages and the program's input, output and command-line arguments. We have successfully reconstructed the overall tool pipeline as a workflow by calling these services from the Taverna Workbench [1].

### 3 Ontology-based Service Discovery

Although web services can be composed into a workflow by simply selecting them from a list, as allowed by the Taverna Workbench, it would be desirable to use a more semantically oriented approach to composing services. To move towards a semantic service environment, over the last couple of years the OWL-S upper ontology for describing web services has been developed. The intention of this effort is

to provide an XML-based ontology which stipulates the basic information that services should expose to the environment in order to facilitate their automatic discovery, invocation, composition and monitoring [20]. This ontology is specified in the OWL Web Ontology Language which provides a language for specifying Description Logic constructs in the syntax of XML and built on top of the RDF data model. For the purposes of discovery, the most important aspect of the OWL-S ontology is the **Service Profile**. This describes the essential capability of the service by characterizing it in functional terms: the inputs a service expects, the outputs it produces, the preconditions that are placed on the service and the effects that the service has. By 'typing' data in this fashion, the ability to define and instantiate 'semantic pipelines' of data through a workflow consisting of a number of distinct services also becomes possible.

The OWL-S ontology is domain-independent: to express concepts of particular domains it is necessary to extend the OWL-S ontology into these domains through the use of additional ontologies. We have developed such an ontology in the domain of NLP (specifically, by extending the OWL-S notion of a **Profile Hierarchy**).<sup>6</sup> Our approach rests on the following assumptions:

1. NLP processors have documents as both input and output.
2. Documents have properties which impose preconditions on the processors and which also record the effects of processing.
3. A specification of the properties of documents, as input/output parameters, induces a classification of NLP processors.

Figure 4 illustrates the case of a part-of-speech tagger, defined so that its input document is marked-up for word and sentence tokens, while its output is a document which is in addition marked-up for parts of speech.



Figure 4: The Tagger class as a function

We can subclass Document class in various ways in order to further refine different classes of processors, as illustrated in in Figure 5.

In order to be able to reason about NLP services, we have developed a broker service, built on top

<sup>6</sup>For more details, see [15].

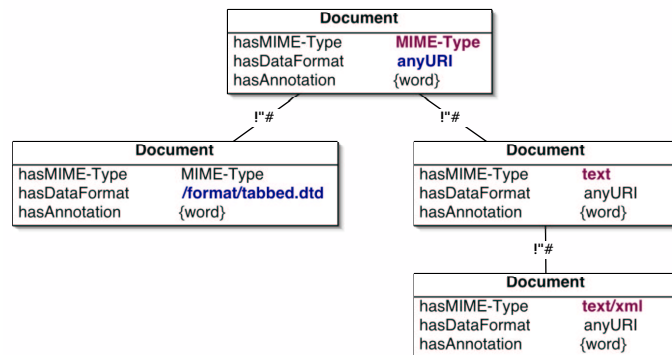


Figure 5: Subclasses of Document

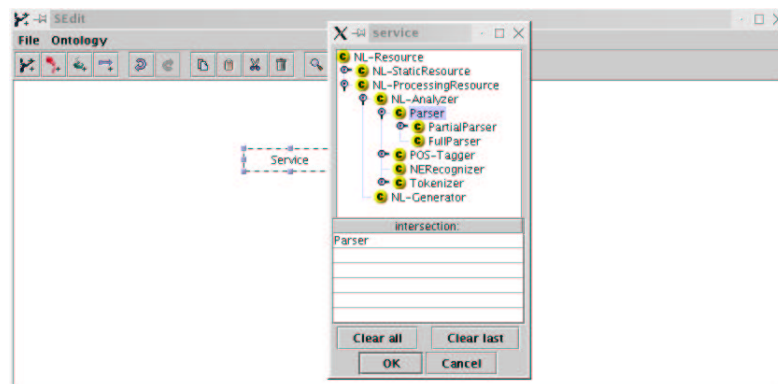


Figure 6: Selecting a Text Mining Tool from the class hierarchy



of the RACER [13] Description Logic engine. This broker maintains a description of the NLP ontology; when it receives service advertisements, described using OWL-S and this domain ontology, it classifies these and stores them as instances of the appropriate class in the hierarchy. On receiving an OWL-S query, it composes a class description from this and then returns, as potential solutions, any service instances of classes subsumed by this description. On the client side, we have developed a prototype composition tool for composing sequences of services and querying the broker. The user is able to specify either the type of processing resource that is needed, or the constraints on the data inputs and outputs to some abstract service (or a combination of both) and the tool constructs the appropriate OWL-S, sends this to the broker and then presents the alternative services to the user. The selection menu offered to the user is shown in Figure 6. Once a user selects one of these, the tool fetches the URL of the service to extract more detailed information about the service, and the composition view is updated accordingly. The composition tool is able to output its configuration file in Scufi format [2], thus enabling the workflow to be saved, and to be enacted within Taverna.

The text processing web services that we described in Section 2 can be classified within the framework of our NLP Profile Hierarchy, and should allow services such as the geoparser to be composed by semantically-based brokering, although the exact details of such a composition are still under investigation.

## 4 Conclusion

In this paper, we have addressed two potential user populations: the scientist who is a potential consumer of text mining applications, but is not interested in the internal make-up of the applications, and the computational linguist, who wishes to construct text mining applications from a combination of in-house and third-party tools. We believe that both groups can benefit from ‘virtualizing’ language processing tools as web services.

One of the key challenges in building on this paradigm is to present the processors in a way that supports robust interoperability. This can only be achieved if the input and output types of processors can be described using a rich vocabulary. We have argued here for an ontology that leverages properties of the class of documents that can be exchanged between the processors; although this does not exhaust all the constraints on interoperability, it does provide a promising basis for capturing the functional core, especially in terms of the data transformations car-

ried out by NLP processors.

Given the huge amount of linguistic information available online, from natural language annotations in databases to the text available on the Web itself, robust and scalable NLP applications are urgently needed for text-mining. By wrapping NLP components as web services, cutting-edge technology can be made readily available for use for real-world problems. In this paper, as a proof-of-concept, we showed how a real-world application could be constructed within a standard workflow environment.

There are a number of issues for future research. For very large documents, the network latency involved in web services could lead to possibly slower results than a single central installation. Furthermore, many text processing components have a heavy statistical component which can make their processing time over large documents slow as well. With the further development of the Grid and its integration with web services, the problems of limited bandwidth and slow processing time should be compensated for. For example, a statistical tagger that invokes high computational costs could have its workload shortened by sending itself as a job to the Grid, and so play a role in a workflow without serving as a bottleneck on performance time. There is already evidence that for some NLP tasks the use of the Grid as opposed to a single system results in improved performance time over very large data sets, and the results get improve as the data set gets larger.[] In this regard, the further development of web services and e-science is likely to play an increasing important role in the development of large-scale, rapid, high-performance text-mining.

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