

# Data Extraction from Web Data Sources

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## Abstract

*This paper provides an explanation of the basic data structures used in a new page analysis technique to create wrappers (data extractors) for the result pages produced by web sites in response to user queries via web page forms. The key structure called a *tpGrid* is a representation of the web page, which is easier to analyse than the raw html code. The analysis looks for repetition patterns of sets of *tagSets*, which are defined in the paper.*

## 1. Introduction

A *web data source* in the context of the current paper is a database backed web server which accepts queries in web page forms and returns the set of result items for that query in a web page. We therefore refer to query

with any web page that displays one or more sets of items that form *collection* objects in the web page. (A collection is a set of items with similar appearance on screen).

The key component of a wrapper for a web data source is the Extractor program. It uses some description of web page format in order to identify and extract data items from the html code representing the web page. Wrapper production therefore requires a *page analysis* phase, before the extractor can be used for the first time. The purpose of page analysis is to create a suitable *page descriptor* for the wrapper to use repeatedly to extract data from future pages from that web site.

An HTML document contains only tags and text items. (Embedded program scripts are not part of the html code, and are removed before page structure analysis begins). A web page can therefore be modelled as the numbered sequence of

```

<HTML><HEAD><TITLE> Spa Guide </TITLE></HEAD>           (S0)
<BODY> <H1> Nunohiki Spa </H1>                        (S1)
  <EM> Location </EM>                                  (S2)
    <P> Kita-Saku district, Nagano pref. Japan </P>    (S3)
    <P> TEL: 0268-67-3467 </P>                        (S4)
  <EM> Open Hours </EM>                               (S5)
    <P> From 10.15 a.m to 11.30 p.m. </P>             (S6)
  <EM> Overview </EM>                                 (S7)
    <P> This facility was established in 1987, and ... <BR> (S8)
    The hot spring is located at the front of ...     (S9)
    <B> "SAWARA", </B>                                (S10)
    on the basement floor ... <BR>                   (S11)
    The effects of this hot spring are ... </P>       (S12)
</BODY></HTML>

```

**Figure 1.** An HTML document, showing textStrings and tagStrings

extraction from query result web pages, but the method of wrapper production and data extraction also works

tagString/textString pairs in its html code, where a textString is anything that is not inside angle brackets,

so it is not a tag. A tagString is the sequence of tags that precede any textString. For example, the html document in Figure 1 shows each textString on a separate line. The lines have been labelled, in Figure 1, with the numbers of the textStrings S0, S1, S2, etc., in order to explain what is meant by a numbered sequence of text items in each html page.

The tagString preceding textString S0 is <HTML><HEAD><TITLE>. The tagString preceding textString S1 is </TITLE></HEAD><BODY><H1>. Each tagString/textString pair has the same number as the textString it contains. The numbered sequence of tagString/textString pairs for the whole HTML document in Figure 1 is shown in Figure 2.

```
</td></tr></table><table valign="top" width="757"
border="0" cellspacing="0" cellpadding="0"><tr><td
valign="top" width="146"><table border="0"
cellspacing="0" cellpadding="0" width="146"><tr><td
width="146"></td></tr><tr><td
align="right" width="146" valign="middle"><a
class=header href="http://www.bl.uk/index.shtml">
```

This tagString precedes a textString in the html document. Ignoring tag attributes (inside the angle brackets) shows that it contains the following tag names:

```
</td></tr></table><table><tr><td><table><tr><td>
```

0	<HTML><HEAD><TITLE>	Spa Guide
1	</TITLE></HEAD><BODY><H1>	Nunohiki Spa
2	</H1><EM>	Location
3	</EM><P>	Kita-Saku district, Nagano pref. Japan
4	</P><P>	TEL: 0268-67-3467
5	</P><EM>	Open Hours
6	</EM><P>	From 10.15 a.m to 11.30 p.m.
7	</P><EM>	Overview
8	</EM><P>	This facility was established in 1987, and ...
9	 	The hot spring is located at the front of ...
10	<B>	"SAWARA",
11	</B>	on the basement floor ...
12	 	The effects of this hot spring are ...

Figure 2. The numbered list of tagString/ textString pairs

All textStrings are visible on the Web Page represented by an HTML document. So a search for repetition in text items in an html document is a suitable way to start the search for collection objects in a web page. Figure 1 shows a document that contains only one record, but the real web page listing spas has a similar record for each of a number of different spas. Each record is displayed on-screen in the same format as the other records, so that the web page structure is understandable to a person reading the page. The format is produced by tags in the html document, and we can discover repeating patterns in the sequence of tagStrings.

The task of searching for repetitive patterns is simplified by observing that *within the local context of tagStrings, the order of tags is not important*. This is an experimental observation. So each tagString is represented by a tagSet. A tagSet is a set of tag names with a count value for each name that occurs in the corresponding tagString.

For example: The tagStrings shown in Figure 2 are too simple to illustrate the idea of tagSets, so a typical tagString from a real web page is now used:

```
<img></td></tr><tr><td><a>
```

Start and end tags (such as td and /td) are treated as different tag names. Counting the number of times each tag name occurs in the tagString produces the tagSet:

```
</td - 2> </tr - 2> </table - 1> <table - 2> <tr - 3>
<td - 3> <img - 1> <a - 1>
```

In order to facilitate tagSet comparisons, we represent each tagSet as a vector of count values. The number of elements in the vector is the number of tag names in the whole html document. There were 41 tag names in the document from which this tagString was taken, so its tagSet is represented as:

```
0,0,0,0,0,0,0,0,2,3,3,1,1,0,0,2,2,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0
```

Which is a 41-element vector of count values. The 41 tag names were:

html, head, title, /title, meta, link, /head, body, table, tr, td, a, img, /a, br, /td, /tr, /table, SPAN, div, /div, form, center, input, /center, hr, strong, /strong, p, font, /font, select, option, /select, b, /span, /FORM, blockquote, /blockquote, /body, /html.



S23: Unidentified flying objects: reports and newspaper cuttings  
 S24: 1972

There are 69 textStrings (and therefore tagSet/textString pairs) in the *whole* document. In the 69 tagSets that occur in these pairs, there are only 12 *different* tagSets, because some of the tagSets are re-used in many different tagSet/textString pairs. The 12 *Distinct TagSets* are shown in Figure 4.

because the next tagSet/textString pair to contain a new tagSet is the one containing textString 67

An *itemSet* is the set of one or more tagSet/textString pairs associated with each distinct tagSet. If we represent each item by its number (in the sequence of numbered pairs) and display the itemSet associated with each of the 12 distinct tagSets shown in Figure 4, we obtain a *tagSet Progression Grid* (tpGrid) shown in Figure 5. This is the abstraction of the html document that we use to analyse page

T0:	1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0
T1:	0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0
T2:	0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0
T3:	0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0
T4:	0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0
T5:	0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0
T7:	0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0
T8:	0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0
T9:	0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0
T10:	0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0
T67:	0, 0, 0, 0, 0, 0, 0, 2, 3, 0, 0, 0, 3, 2, 2, 1, 1, 3, 2, 0, 0, 0, 0, 0
T68:	0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0

**Figure 4.** The Distinct TagSets in Figure 3's html document

Each row in that table represents a different tagSet. Each tagSet is a vector of 24 count values, for each of the 24 tag names found in the whole of the html document. The 24 tag names are: html, head, meta, /head, body, table, tbody, tr, td, p, b, /p, a, img, /a, font, /font, /td, /tr, /tbody, /table, /b, /body, /html.

structure for repetitive features. It is much easier to see repetition in this grid structure than in the raw html code.

In order to avoid confusion it is important to distinguish the meaning of figures 4 and 5. Figure 4 shows all the Distinct TagSets in the html document.

T0:	0
T1:	1
T2:	2
T3:	3
T4:	4
T5:	5, 6
T7:	7
T8:	8, 11, 14, 17, 20, 23, 26, 29, 32, 35, 38, 41, 44, 47, 50, 53, 56, 59, 62, 65
T9:	9, 12, 15, 18, 21, 24, 27, 30, 33, 36, 39, 42, 45, 48, 51, 54, 57, 60, 63, 66
T10:	10, 13, 16, 19, 22, 25, 28, 31, 34, 37, 40, 43, 46, 49, 52, 55, 58, 61, 64
T67:	67
T68:	68, 69

**Figure 5.** A tpGrid. It shows the itemSet associated with each Distinct TagSet in Fig 4

Each row is numbered with the number of the tagSet/textString pair in which the tagSet was first used in the html code. TagSet10 is the third row from the end. All tagSet/textString pairs from 11 to 66 (in the sequence of numbered pairs that represent the html code) must contain one of the tagSets already used,

Figure 5, on the other hand, does not show any tagSets. Although the rows in Figure 5 are labelled with tagSet numbers, which *correspond* to those for the same rows in Figure 4, each row in Figure 5 shows the *itemSet* associated with each tagSet.

The *row cluster* T8, T9, T10 in Figure 5 reveals the position of the collection of 3-field result records in the

web page. It also provides all the information an extractor program needs to accurately find and extract the data item for each field in each result record.

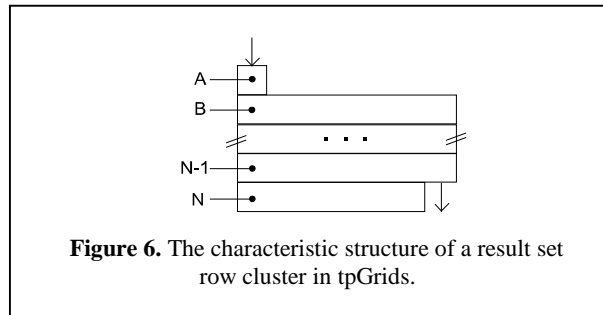
Figure 5 contains all the text items visible on the web page. Each text item is identified by its position number in sequential order. Rows T8, T9, T10 contain items 8 to 66. This cluster of long rows corresponds to the set of 3-field results in the web page. By constructing the tpGrid we have found the position of the results set in the web page, discovered that result records contain three fields, identified distinct tagSets that mark the start and end of the results section of the html document, and found other tagSets that uniquely label each field's data items so that accurate extraction of data items into correctly formatted records is ensured.

The way to discover and interpret patterns in a tpGrid (e.g Figure 5) is to follow the **TRAIL** of item numbers. If consecutive item numbers were linked with a line the result would be the Trail. It shows the order in which tagSets are used in the web page by following the sequence of numbered tagSet/textString pairs that correspond to the html document. *For example*, in Figure 5, tagSet T0 occurs only once in the whole document: before textString 0. Similarly we see that tagSets T1, T2, T3, T4, T7 and T67 each appear only once in the entire html document (preceding textStrings 1, 2, 3, 4, 7 and 67, respectively). TagSet T5, in contrast, has an itemSet containing two items: 5 and 6. This means that the two tagSet/textString pairs numbered 5 and 6 both contain the same tagSet, T5.

TagSets T8, T9 and T10 have large itemSets. A large itemSet is shown as a long row of integers. Each integer represents by number a numbered tagSet/textString pair that contains the same tagSet as all the other items in the row. TagSet T8 occurs before twenty textStrings, namely textStrings 8, 11, 14, 17, 20, 23, etc. Notice that the sequence of item numbers forms an arithmetic progression with common difference 3. This is because result records contain three fields. **The Trail** follows repeated vertical sequences of three items while it is inside the row cluster. TextStrings 8 to 66 on the web page use only tagSets 8,9 and 10, and the use of those three tagSets is *cyclic*.

Each of the three fields in each record has a distinct tagSet to identify items belonging to that field. However, tagSet T8 does not identify first field items, as might be supposed. Instead, the row cluster representing a set of results has a characteristic structure in tpGrids, which is shown in Figure 6, where Trail entry and exit points are shown by arrows.

Rows T7 to T10 in Figure 5 are an example. The tagSet before the first record (row **A** in figure 6) is different from the tagSet (row **N**) before each



**Figure 6.** The characteristic structure of a result set row cluster in tpGrids.

subsequent record in the result set. **A** is tagSet T7 and **N** is tagSet T10 in figure 5. So item 7 is the first field data for the first record and item 10 is the first field of the second record. The first field items of all the other records are in row T10. The reason for this structure is that the tagSet before the whole set of results uses different tags from that which just separates records within the collection of results.

### 3. The Data Extraction Algorithm

The tpGrid (Figure 5) provides the information needed by an extractor (wrapper) program to extract data from result pages from this web site. The extractor program proceeds as follows:

1. Search for tagSet T7 which occurs only once in the html document. It marks the start of the result set. TagSet T7 is immediately followed by the data item for the first field of the first record. Records have three fields.
2. TagSet T67 follows the last data item in the result set. It can be used either to isolate the results section from the html document, or just as a stopcode: stop extracting data items when tagSet T67 is encountered.
3. After the first field of the first record, a *field2* data item is preceded by tagSet T8; a *field3* data item is preceded by tagSet T9, and a *field1* item is preceded by tagSet T10. Each data item is, in effect, labelled in the html code by the tagSet that precedes it. In result sets whose field order can vary between records, this unique labelling is a significant benefit. And also when field values are missing from result records.
4. The extractor continues to extract data as long as the specified tagSets are found. Pages which contain no data (because the query produced no results) cause no problem because the tagSet labels for data items will not be present either. So the extractor recognizes an empty result set.

## 4. Rearranging tpGrids

The tpGrid representing a web page allows collections of result records to be identified as blocks of long rows (i.e. row clusters) with the structure shown in Figure 6. But sometimes items or whole rows in a tpGrid are *displaced* from their correct position in the row cluster to which they belong. A row is displaced if its tagSet happens to be used somewhere else in the web page, as well as in the part of the page containing the results. Individual items are displaced if an *insignificant* variation in their set of preceding tags gives them a different tagSet from other items in the same field position in records. Such displaced entities in the tpGrid can be automatically moved to their correct position (in order to reveal the repetition pattern). The process involves Trail-following and tagSet comparison, as now illustrated by means of an example.

An IEEE Search results web page is shown in Figure7 (from <http://www.computer.org>). Each result record is shown in a white rectangle and contains a fairly large number of potential field items. The tpGrid for this page, before repositioning displaced rows is shown in Figure8, and after rearrangement in Figure 9.

The ten result records visible in Figure 7 are shown in Figure 9 as the 10-column row cluster between rows T41 and T51. By following the Trail through this cluster we observe that each record has eleven fields, because of the eleven items in each column in the cluster of long rows.

Item 127 is out of place. It should be in the space above it, in tagSet T50's row of items. The eleven fields in other columns associate a particular tagSet with each different field. So we expected item 127 to have tagSet T50, but instead it has new tagSet T127.

Comparing tagSets T50 and T127:

```
50:0,0,0,0,0,0,0,0,0,0,0,0,2,2,2,2,2,0,0,0,...,0,0,0
127:0,0,0,0,0,0,0,0,0,0,0,0,2,1,1,1,2,0,0,0,...,0,0,0
```

The 3 extra tags, absent from tagSet 127 are: a, img, /a, so there is an image hyperlink missing. This absence can be seen in Figure 7: in the third record (white rectangle) from the end, the usual row of four icons has only three. Missing images and hyperlinks are common variations found between records in result sets, so they are automatically recognised as *insignificant variation* between tagSets.

The BigBook standard web page from the RISE repository [11] produces a tpGrid that is structurally analogous to Figure 9, although the Web Page looks very different from Figure 7. Our analysis of the BigBook page structure can be seen at web site [10].

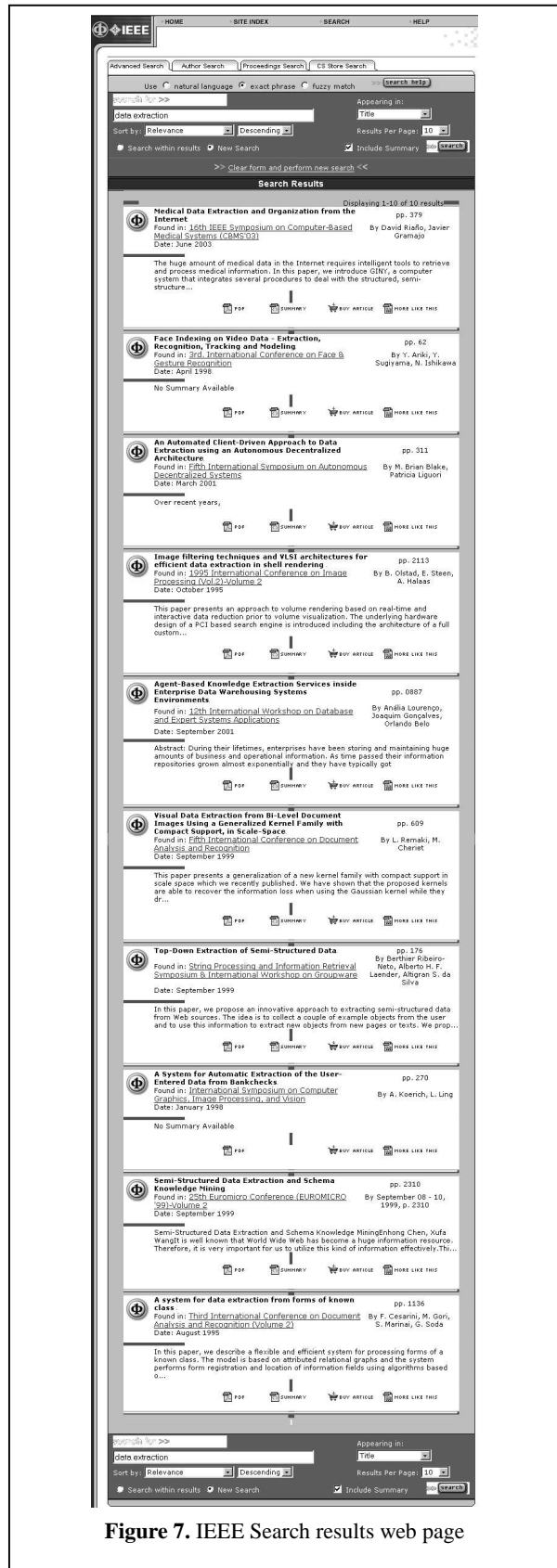


Figure 7. IEEE Search results web page

```

T0: 0
T1: 1
T2: 2,3,4, 27, 173
T5: 5
T6: 6, 152
T7: 7,8,9,10, 13,14,15,16, 18, 153,154,155,156, 159,160,161,162,164
T11: 11, 157
T12: 12, 158
T17: 17, 163
T19: 19, 165
T20: 20, 166
T21: 21,22,23,24,25, 167,168,169,170,171
T26: 26, 172
T28: 28, 174
T29: 29
T30: 30
T31: 31
T32: 32
T33: 33
T34: 34
T35: 35
T36: 36
T37: 37,38,46,48,57,59,68,70,79,81,90,92,101,103,112,114,123,125,134,136,145,147
T39: 39
T40: 40
T41: 41,52,63,74,85,96,107,118,129,140
T42: 42,45,53,56,64,67,75,78,86,89,97,100,108,111,119,122,130,133,141,144
T43: 43,54,65,76,87,98,109,120,131,142,176
T44: 44,55,66,77,88,99,110,121,132,143

T47: 47,58,69,80,91,102,113,124,135,146

T49: 49,60,71,82,93,104,115,126,137,148
T50: 50,61,72,83,94,105,116,138,149
T51: 51,62,73,84,95,106,117,128,139
T127: 127
T150: 150
T151: 151
T175: 175
T177: 177
T178: 178

```

**Figure 8.** The tpGrid before rearrangement, for the web page shown in Figure 7

## 5. Terminology

This section defines some of the key terminology associated with the ideas introduced during the discussion of examples above.

A **Pair Sequence** is a numbered sequence of tagSet/textString pairs. Each tagSet summarises a corresponding tagString in the html document. The Pair Sequence for a web page is an indexed version of the html document representing that page. It contains all the information in the html code (including the final tagString that has no textString after it, but is followed instead by the end of the document/file). It allow Ti/Si pairs to be identified by number, and accessed by number when required. So the Pair Sequence is a lookup table, accessed by item number. The reason it contains all the html code in the html document is that

each tagSet summarises a corresponding tagString, so each numbered pair can be seen as a tuple: [ < tagString || tagSet >, textString ] whose first field is a tagSet and its corresponding tagString. So information from the raw html at any point in the document can be accessed by position number if or when required. The page analysis process described in the current paper uses only the tagSet part of the embedded < tagString || tagSet > tuple.

**Potential data items** (PDIs) are the numbered items in the Pair Sequence. A basic goal of page analysis is to discover which of the PDIs contain data items. Data is contained in tag attributes that are hyperlinks to files, as well as the textString. The repetition of tagSets reveals collections of PDI records whose fields are each a [ < tagString || tagSet >, textString ] tuple, rather than just a textString. So a sequence of field

```

T0: 0
T1: 1
T2: 2,3,4,27,173
T5: 5
T6: 6,152
T7: 7,8,9,10,13,14,15,16,18,153,154,155,156,159,160,161,162,164
T11: 11,157
T12: 12,158
T17: 17,163
T19: 19,165
T20: 20,166
T21: 21,22,23,24,25,167,168,169,170,171
T26: 26,172
T28: 28,174
T29: 29
T30: 30
T31: 31
T32: 32
T33: 33
T34: 34
T35: 35
T36: 36
T37: 37,38
T39: 39
T40: 40
T41: 41,52,63,74,85,96,107,118,129,140
T42: 42,53,64,75,86,97,108,119,122,141
T43: 43,54,65,76,87,98,109,120,131,142
T44: 44,55,66,77,88,99,110,121,132,143
T42: 45,56,67,78,89,100,111,122,130,144
T37: 46,57,68,79,90,101,112,123,125,145
T47: 47,58,69,80,91,102,113,124,135,146
T37: 48,59,70,81,92,103,114,125,134,147
T49: 49,60,71,82,93,104,115,126,137,148
T50: 50,61,72,83,94,105,116,138,149
T51: 51,62,73,84,95,106,117,128,139
T127: 127
T150: 150
T151: 151
T175: 175
T177: 177
T178: 178

```

**Figure 9.** The tpGrid after moving displaced rows, for the web page shown in Figure 7

items may be identified in the tagString preceding each textString. Each single field in the discovered repetition pattern may thus be subsequently expanded into several fields (the extra fields are URLs extracted from tag attributes in the tagString of the PDI).

## 6. Automatic page format change detection

One of the key problems that has limited interest in using web data sources is their brittleness of access. A web site only has to change the layout of its results page and a wrapper will no longer extract data correctly.

A system is needed that *monitors* page format, so that any change is detected *before* a query to the site fails. This is easily achieved with tpGrids, because a grid is a fingerprint for page format. Therefore a query

sent repeatedly to a web site will generate the same structures in the tpGrid for the results page. So by monitoring the grids from such test queries it is possible to detect page format changes as soon as they occur, and also to distinguish between changes that will affect data extraction and those involving other parts of the page. This fingerprint for page format is a valuable resource to ensure reliable data access.

## 7. Discussion and Experimental Evaluation

The page analysis technique has been applied to a large number of web site result pages in order to investigate its effectiveness. Examples of the range of different pages successfully wrapped by the page analyser can be seen at our web site. Our ultimate goal is fully automatic page analysis (and hence wrapper



production) because this enables new applications for autonomous agents. But many of the current uses for wrappers (such as data integration systems, shopping agents and recommender systems, for example) probably *want* some minimal human involvement in order to provide confidence that completely accurate data extraction is achieved but also to specify the schema for the extracted data. Not all available fields are wanted. They could be extracted and then discarded or ignored later. Or there may be a wish to divide one text string into more than one field during extraction. (This can alternatively be done later by processing the extracted database table).

If the page analysis process is used with human input to resolve ambiguities then it is effective at wrapping the majority of web pages seen so far. (We exclude, of course, plain text documents studied in NLP Information Extraction research. Our system looks for repetition patterns in html tags). Pages with few results provide limited repetition information. This problem can be solved by comparing the tpGrids for two or more web pages from the same web site (to find the results by distinguishing constant from variable page components).

Nested iterations were discussed briefly in [12]. Further examples are shown in our web site [10]. Pages whose data set has a tree structure have a tpGrid results section that is a nested version of Figure 6.

Our work is continuing by examining further web site results pages, to discover and resolve any problems. A comparison with previous work by processing the 'standard' web results pages in the RISE repository [11] will be published in due course. We are investigating the use of various knowledge sources in order to automate the process of resolving ambiguities. The use of repetition patterns of tagSets alone has been very effective at discovering page structure and identifying tagSet labels for data items, to be used by the extractor. But there is a lot of other information that can be used as well, to advance the aim of fully automatic wrapper production for *any* site. Many web sites have completely regular result sets (in terms of tagSet repetition) even though some fields may be missing. Such sites are automatically wrapped just by tpgrid analysis.

## 8. Related Work

Previously published work on the task of extracting data from *documents*, to create database tables, is divided into two separate research areas: Information Extraction and Wrapper Production. The former applies to plain text documents (rather than web pages) and uses Natural Language Processing techniques to

obtain some understanding of the document in order to identify and extract substrings as data items. That is very different from Wrapper Production, which uses *the html framework* to divide up the document and recognise parts that are required data. This makes wrapper production *language independent* since the techniques apply to html rather than the language displayed on screen.

Previous work on wrapper production has included a lot of work using Machine Learning algorithms on html pages that have been previously annotated by hand to identify and describe the data items embedded in the document. This approach is clearly not an automatic technique, where a program finds the data and discovers its structure. Other researchers have tried to remedy that deficiency. Searching for repetition patterns has been tried in several ways. Repeating substrings in the html document have been sought, repeating subtrees in the html parse tree, and also repeating individual tags that recur at approximately equal intervals within the document character string. These methods experienced difficulties because of the ways they represented the document. In contrast, the tpGrid data structure represents a web page in terms of potential data objects and in a way that immediately reveals relevant features of page structure. This abstraction is a much easier thing to analyse than the document itself, and its item numbering system makes it a kind of index into parts of the original web page so that details can be obtained if required during analysis.

Another method [2] previously described for automatic wrapper production compared two or more query result pages from a web site. The purpose was to recognise constant parts that occur in all result pages from a site. Hence the other parts may contain data.

Domain-specific ontologies have also been used to support wrapper production techniques. These are specific to the kind of data to be extracted from web pages. The approach differs from other wrapper production strategies in considering the text of the data substrings rather than the html framework. It is thus language and domain dependent.

Our approach may compare web pages in order to distinguish between iterations that contain data and others that are part of the page structure. A simple ontology *for web result pages*, rather than data, is helpful to recognise parts of the web page such as the link to fetch the next page-full of query results.

## 9. Conclusions

Creating an extractor program for the result pages from a particular web site involves a page structure analysis process. An analyser program examines one or more

pages from the web site and produces a form of page description to be used later by the extractor program on any query result pages obtained from that site.

The tpGrid is a useful object for analysis, supplementing information a program can obtain by directly examining the html code for the page. The grid finds data in the page, discovers its record structure and also identifies the tagSet 'labels' that an extractor program can use to recognise each type of data item to extract from other results pages from the same source.

The tpGrid is a fingerprint for the page structure of a particular web site. So it is easy to detect page format changes before they cause a wrapper to fail during extraction.

Since the method operates on repetition of items observed in the results page, it will not find data on pages that display few result items. For these, either a test query that generates more data is needed, or else the comparison of two or more results pages to distinguish between constant and variable parts of result pages.

The method of page analysis discussed in this paper is quick and simple, showing potential for use by 'robot' programs exploring the 'hidden web'. Techniques for automatic form-filling were discussed in [6].

## 10. References

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