Web Search using IR

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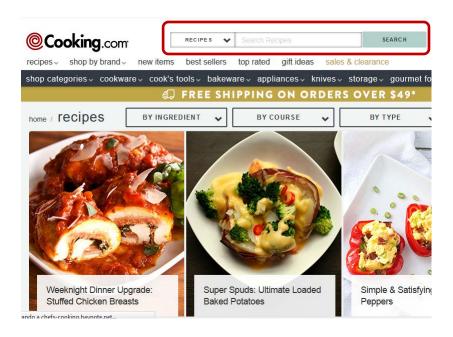
References

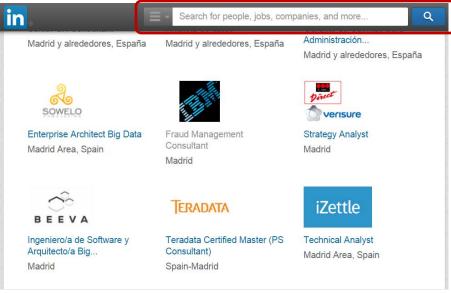
Introduction

The Internet

- 4,7 billion pages !!
- Lots of Multimedia Data







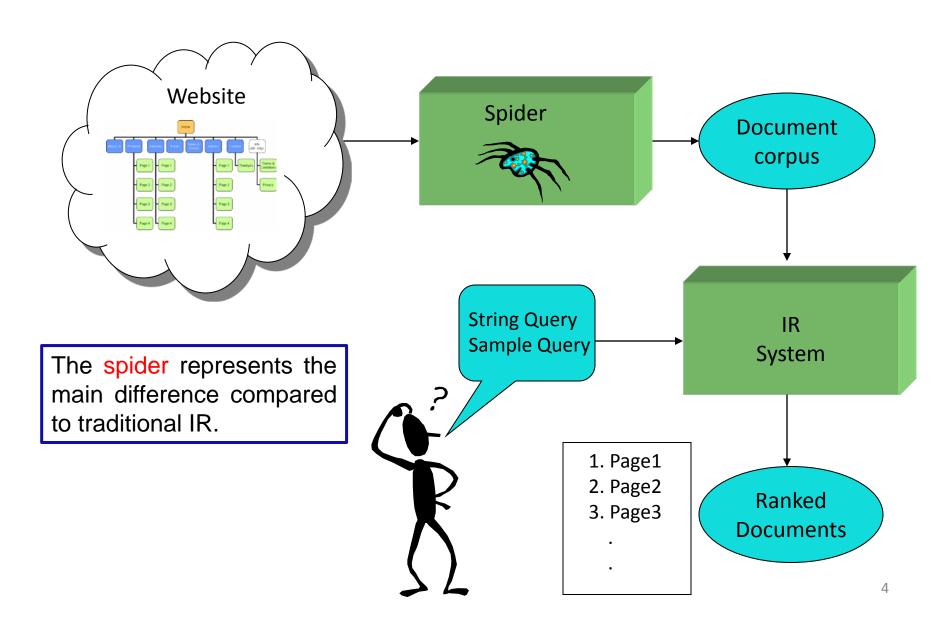
Too many documents for performing classical Information Retrieval

- Indexing and Page Rank
- Specific web searchers just constraint the URL.

Constraint the document number

- Specific topics
- Perform better IR models
- Allow Example Query

Structure



Web Crawling + HTML processing

How do we download the Web pages?

How do we obtain the:

- Relevant text
- Metadata

from the HTML documents?

Web Crawling

We must download our corpus from the Internet using a Spider or WebCrawler.

A **Spider is a bot** that:

Travels through WebPages downloading and procesing their content.

Parameters:

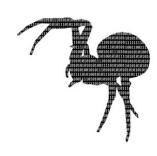
- Starting URLs
- Allowed domains
- Link-following protocol
- Rules for scraping.
- Document Processing function.



- Not good enough for large scale downloads.
- Ended up using the Scrapy framework.







Web Crawling

Scrapy is a scraping and web crawling framework, used to crawl websites and extract structured data from their pages.

```
111 class all recipes spider(CrawlSpider):
       name = 'all recipes'
112
                                                                 URL properties
       allowed domains = ["allrecipes.com"] # Allowed domains
113
       start urls = ['http://allrecipes.com'] # urls from which the
114
       rules = [Rule(SgmlLinkExtractor(allow=[r'.*(recipe).*']), #\intracts all pages in the folder recepi
115
            callback='parse doc', follow=True),
116
117
                                                                      Crawling Rules
118
       #.* means any number of chars
       rules = [Rule(SgmlLinkExtractor(allow=[r'']), # Stracts all
119
             follow=True). # To follow more shit
120
121
122
       def parse doc(self, response):
           filename = response.url # Writes a file with the name of the document
123
124
           filename = filename[7:] # Eliminate the http:// part
           if (filename[-1] == '/'): # If the end of the URL is j
125
               filename = filename + "index" # Give a name to the
126
127
           if not os.path.exists(os.path.dirname(filename)): # If
                                                                   e folder of this does not exist
128
               os.makedirs(os.path.dirname(filename))
                                                            # Make the folder
           with open(filename + ".html", 'wb') as f:
129
130
               f.write(response.body)
```







Corpus Preprocessing

Once we have folders structures containing **HTML documents** we have to:

- Read all possible documents, need a «walker» program.
- Remove non-recipe documents.
- Remove near duplicates.
 - Due to JavaScript generating pages





An HTML document has:

- Plain text.
- HTML markups
- Scripts: JavaScript, PHP...
- Syles: CSS
- Hyperlinks

We need to **extract the relevant information**:

- Remove Scripts and Styles
- Remove the markups.
- Get Relevant text

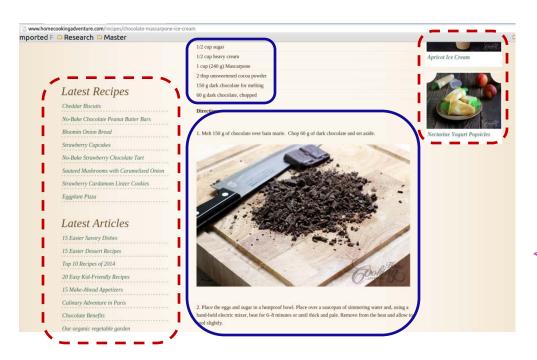
Beautiful Soup!!

```
class="categories">Sodium
              class="units"><span id="lblNutrientValue">434
              class="nutrition-rating">
                      id="divNutrientGradient" class="nutri
                      class="nutrition-rating-img">
                  class="percentages">17%
           * Percent Daily Values are based on a 2,000 calorie diet.
   <a href="javascript:void(0);" id="nutritionSeeMore" class="rr</pre>
   <div id="eshaLinkSummary"><a id="eshaLink" rel="nofollow" tar</pre>
</div>
\textbf{<div id} = \texttt{"nutritionDetail" style} = \texttt{"display:none;" class} = \texttt{"bottomLaye}
   <!-- DETAILED NUTRITION -->
   <div id="nutritiontable" class="nutrDetWrap" itemprop="nutrit</pre>
       <h2>Nutritional Information</h2>
       <span id="lblTitle">Amazing Slow (
           Serving Size: <span itemprop="servingSize">1/12 of a
           Servings Per Recipe: 12<br />
           <strong>Amount Per Serving</strong><br />
           Calories: <strong><span id="litCalories" itemprop="ca
           Calories from Fat: <strong>102</strong>
       class="medBorder">
               <span class="right">% Daily Value *</span>
           i>
              <span class="left"><span></span>Total Fat</span>
```

Corpus Preprocessing

BeautifulSoup library for HTML processing

- Remove Scripts and Styles
- Reference parts of the HTML
- Get Plain texts
- We still have to get our hands into the HTML code !!



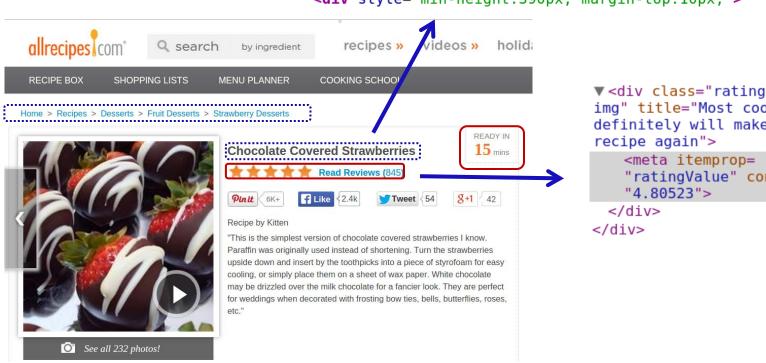
```
<div id="ingredients bg">
               <span class="ingheading">Makes about 4 servir
         </div>
         <div id="preparation">
         <div class="title" >Directions</div>
         <div itemprop="instructions">
         <0l>
   <span>
      Preheat oven to 400F (200C) and line a baking sheet with parch
      Beat the eggs with oil and set aside. Place the flour on a pla
        <span itemprop="servingSize"><b>1 Chicken Strip</b></span> - Cal
  </div>
           <div style="height: 7px; font-size: 7px; margin-bottom:10px;</pre>
<div id="posted_comments" style="margin-top:20px;">
                             <div class="comment top">
     On February 03, 2014 at 08:19 pm, <span><a href="http://cookingact
   <div class="comment middle">
     I LOOOVE how crisp and delicious these look!! And, of course, that
   <div class="comment bottom">
   </div>
```

Corpus Preprocessing

Get **Metadata** Information for Ranking:

- KeyWords
- **Ratings**

```
<div id="recipe title" ><h1 itemprop="name">Baked Cornflake Crus
<meta itemprop="author" content="Home Cooking Adventures" />
<div id="main ima">
            <img src="/images/recipes/baked-cornflake-crusted-ch</pre>
<div style="min-height:390px; margin-top:10px;">
```



▼<div class="rating-starsima" title="Most cooks definitely will make this "ratingValue" content=

IR system

How do we preprocess the Plain text?

Represent documents as an TF-IDF vector?

What Similarity Measures do you use?

Text Preprocessing

Once we have relevant plain text, we use the **NLTK** Python library for preprocessing

- Tokenization
- Lower Case
- Remove non-alphanumeric
- Remove stopwords
- Steamming (Snowball)

String Query!

```
def doc tokeniz(document):
    # Gets the tokens of the document. The tokens are words.
    tokens = word tokenize(document)
    return tokens
def doc lowercase (document):
    # Transforms a list of words (document) into lowercase
    low text = [w.lower() for w in document]
    return low text
def doc_rem_stopwords(document):
    # Removes stopwords obtained from the nltk english corpus.
    stopwords en = stopwords.words('english')
    clean text = [word for word in document if not word in stopwords en]
    return clean text
def doc stem(document):
    # Performs the english steaming of the words
    stemmer = SnowballStemmer('english')
    steammed text = [stemmer.stem(word) for word in document]
    return steammed text
def doc rem punctuation(document):
    # Removes punctuation marks
    clean text = [w for w in document if w.isalnum()]
    return clean text
```

```
I'm feeling like CHocolate with strawberry chickeN salad Flavour. I love chicken!
['I', "'m", 'feeling', 'like', 'CHocolate', 'with', 'strawberry', 'chickeN', 'salad', 'Flavour', '.', 'I', 'love', 'chicken', '!']
['i', "'m", 'feeling', 'like', 'chocolate', 'with', 'strawberry', 'chicken', 'salad', 'flavour', '.', 'i', 'love', 'chicken', '!']
['i', 'feeling', 'like', 'chocolate', 'with', 'strawberry', 'chicken', 'salad', 'flavour', 'i', 'love', 'chicken']
['feeling', 'like', 'chocolate', 'strawberry', 'chicken', 'salad', 'flavour', 'love', 'chicken']
[u'feel', u'like', u'chocol', u'strawberri', u'chicken', u'salad', u'flavour', u'love', u'chicken']
```

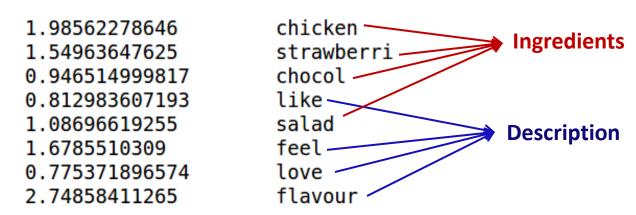
Vector Representation

We represent each HTML document as a vector of float numbers.

- We perform the TF-IDF of every HTML document
- $TF IDF_{(t,d,c)} = TF_{(t,d)} \cdot IDF_{(t,c)}$
 - $TF_{(t,d)}$: Term Frecuency (BoW). Frecuency of term 't' in the document 'd'
 - $IDF_{(t,c)}$: Inverse Document Frecuency of term 't' in corpus 'c'

$$TF_{(t,d)} = \frac{\text{Number of terms 't' in document 'd'}}{\text{Total number of terms in 'd'}} \quad IDF_{(t,c)} = \log_2\left(\frac{\text{Total number of documents in 'c'}}{\text{Number of documents that contain 't'}}\right)$$

String Query!



Similarity measures

Given a **plain text Query D**, we need to:

- Get its TF-IDF vector.
- Compare it with the whole corpus according to a Similarity Measure.

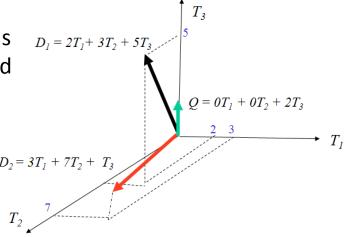
Programming Difficulty:

- Every document has a different set of words $D_1 = 2T_1 + 3T_2 + 5T_3$
- Need to find the words in commun and their associate TF-IDF values.

Similarity Measures:

- Euclidean Distance
- Cosine similarity

Good for Example Queries! Bad for short String Queries and Keywords!



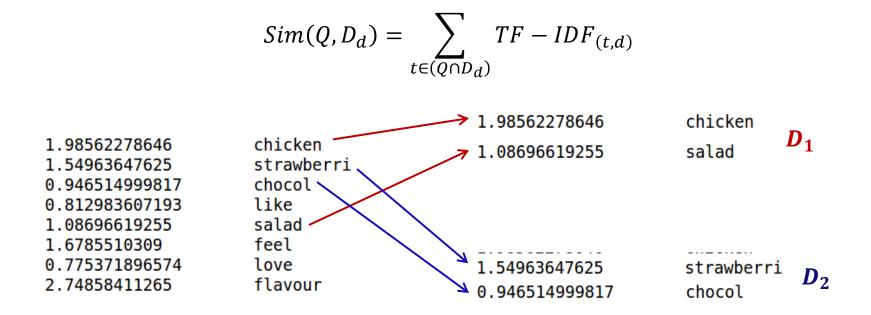
Simmilarity for short Queries

For short Queries, the comparasing between the Quert TF-IDF does not make sense

- Cannot stract real frecuency profile of query.
- Most values are zero and they do not repeat.

Idea!!

Add up the TF-IDF values of the commun words with the query.



Ranking

The BEST similarity measure depends on the properties of the query:

- For Example Queries:
 - ✓ Average ranking of Ecuclidean Distance + Cosine Distance rankings
- For **String Queries**:
 - ✓ Sum of the TFIDF values of the common words.
- For keywords:
 - ✓ Number of words in commun.
- For **Stars Ranking**:
 - ✓ Rule-Based Constraint

Experiments and Results

What is your Corpus made of?

Are your Results any Good?

What Similarity Measures do you use?

Experiments

Corpus of about 1000 recipes!! Not too big to avoid very close recipes!

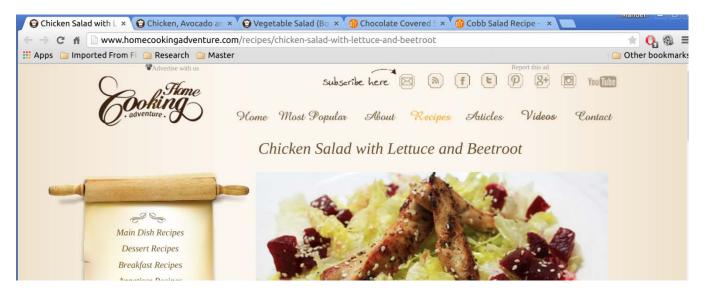
- www.homecookingadventure.com
- <u>allrecipes.com</u>

System accepts:

- Plain text queries
- URLs as a query.
 It downloads and preprocess the HTML document.

The System then opens up a browser with the top 5/10/20... recipes !!

Using the «webbrowser» library



Demonstration Time!

Give me some ingredients!

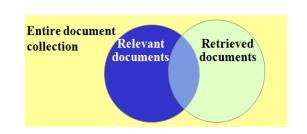
Try to be nice!



Results and Evaluation

Results were evaluated using only **MAP** with K = 5

- Mean Average Precision
- Using Explicit Relevance Feedback

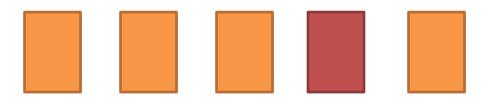


Our corpus is dynamic and unlabeled:

It's costly to perform Recall or Cumulative Gain.

$$Precisiont \ at \ document \ t = \frac{Number \ of \ relevant \ pages \ retrieved}{Total \ number \ of \ pages \ retrieved}$$

On average they follow this structure:



Average Precision =
$$\frac{1}{4}(1+1+1+0.8) = 0.95$$

Conclusions

You can make your own specific corpus!!

- About anything you want !
- In one single day!
- For FREE!

The System:

- Retrives good HTML documents.
- Performs more accuarate Web searchs than Google jejej
- Constrains the search using specific inner web data
- Can use example queries.

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- [2] WU, Ho Chung, et al. Interpreting tf-idf term weights as making relevance decisions. *ACM Transactions on Information Systems (TOIS)*, 2008, vol. 26, no 3, p. 13.
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Thank you!!