

PERSONALIZED BOOKMARK ASSISTANT

Automated Classification of Bookmarks

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INTRODUCTION

PROBLEM DEFINITION

- Users keep track of the valuable Web sites they have visited and where they would like to return, which is done using a personal list of URLs known as bookmarks.
- Well organized bookmarks facilitate user's work and constitute his or her personal information space.
- The current Web browsers offer very limited functionality for the bookmark management, mainly only manual sorting bookmarks into manually created folders.

GOAL

- To Design *personal bookmark assistant*, which organizes bookmarks in an *automated way*
- Bookmark's organizing will be adaptive and incremental, enabling smooth additions of new bookmarks to the recent structure.
- To make the system more *personal* and *assistant*
- System should be able to recommend to a user other Web hyperlinks, which are similar to the already organized bookmarks
- The bookmark personal assistant will classify the bookmarks in the domains:

NEWS	BLOGS	ENTERTAINMENT	SONGS	EDUCATIONAL	SPORTS	OTHERS

which are popular among web surfers.

OBJECTIVES

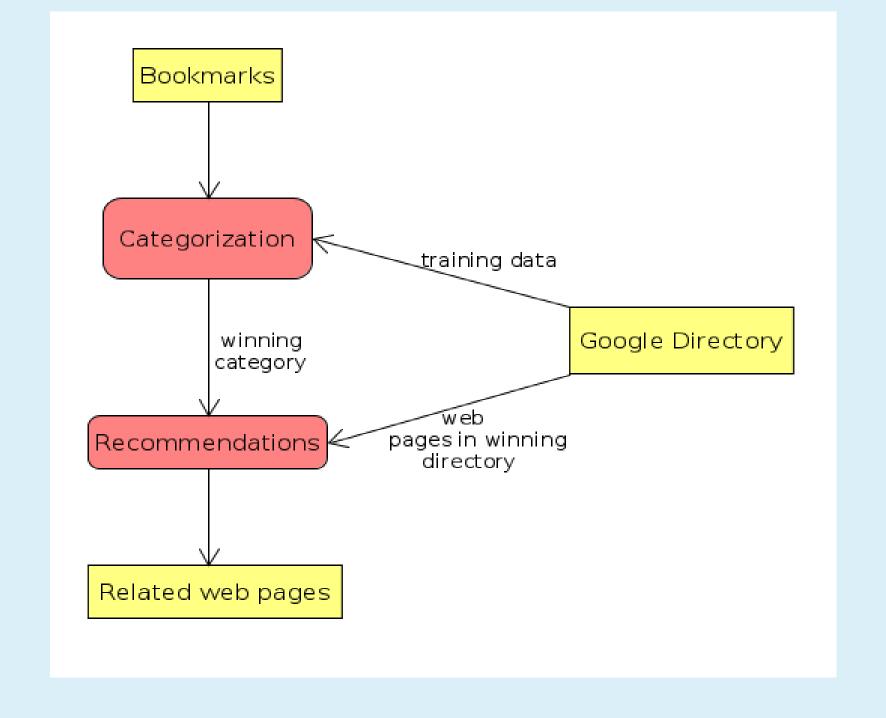
- To develop a personalized bookmark assistant
- To classify the URLS based on the domains: news, blogs, videos, songs, educational and sports
- To be able to recommend to a user other Web hyperlinks, which are similar to the already organized bookmarks
- To make the experience more user-friendly

ARCHITECTURE

TECHNICAL DETAILS

- Techniques Used-
 - ☐ Supervised Machine Learning Approach Textual Classification
 - Natural Language Processing (NLP)
 - Web Crawling
- Languages Used: Python, HTML, CSS, JavaScript, Json
- Library used: Natural Language ToolKit (NLTK)
- Dataset Used : DMOZ (Google Open directory project)
- *Memory Used*: 332 MB by dataset

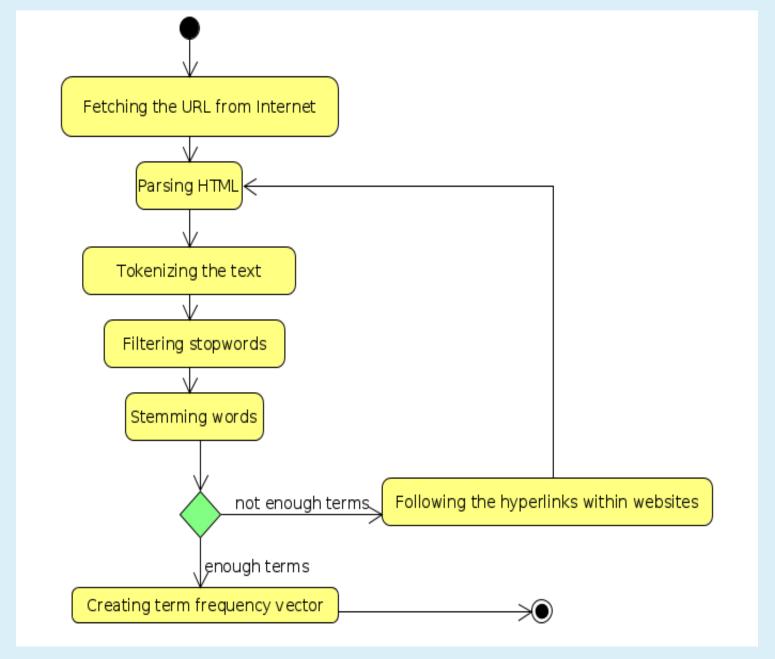
TRAINING OF DATA



TRAINING DATA AND PRE-CLASSIFICATION

- Dataset used: DMOZ Google Open Directory Project
- DMOZ contains URL's of various website in an categorical order
- Domain For our Project is:

NEWS	BLOGS	ENTERTAINMENT	SONGS	EDUCATIONAL	SPORTS	OTHERS



Steps in Pre-Classification

Step 1: HTML parsing HTML parser omits useless parts of the code and extracts only the following: Title Meta Keyword Tag Meta Description Tag Normal Text

☐ Links to other pages which are stored at same serve

Step 2: Tokenisation

- ☐ Extracting words from the text
- ☐ For This we will use nltk.tokenizer

Input: Friends, Romans, Countrymen, lend me your ears;
Output: Friends Romans Countrymen lend me your ears

Step 3: Stop word deletion

- Words that are not beneficial for information retrieval are removed.
- ☐ General English stop words are removed like "the", "of" etc.
- ☐ For This we will use nltk.stopwords

Sample text with Stop	Without Stop Words		
Words			
GeeksforGeeks – A Computer	GeeksforGeeks , Computer Science,		
Science Portal for Geeks	Portal ,Geeks		
Can listening be exhausting?	Listening, Exhausting		
I like reading, so I read	Like, Reading, read		

Step 4: Stemming

- ☐ A program that reduces word forms to a canonical form
- ☐ This is done by removal of the various suffixes like -ed, -ing, -ion, -ions etc. to leave the single term.
- ☐ For this we used porter stemmer
- For instance:

"connect", "connected", "connection", "connections" can be stemmed to the single term "connect".

Step 5: Term Frequency-Inverse Document Frequency (TF-IDF)

- *TF*:
- Measures how frequently a term occurs in a document.
- □ TF(t) = (Number of times term t appears in a document) / (Total number of terms in the document).
- IDF:
- Measures how important a term is.
- \square IDF(t) = log_e(Total number of documents / Number of documents with term t in it).

TESTING DATA

For Testing of data:

- Step 1: Fetch URL
- Step 2: Web Document Indexing similar to training set
- Step 3: Centroid based classification model to determine the category
- Step 4: Updating the dataset with new word vectors from test data

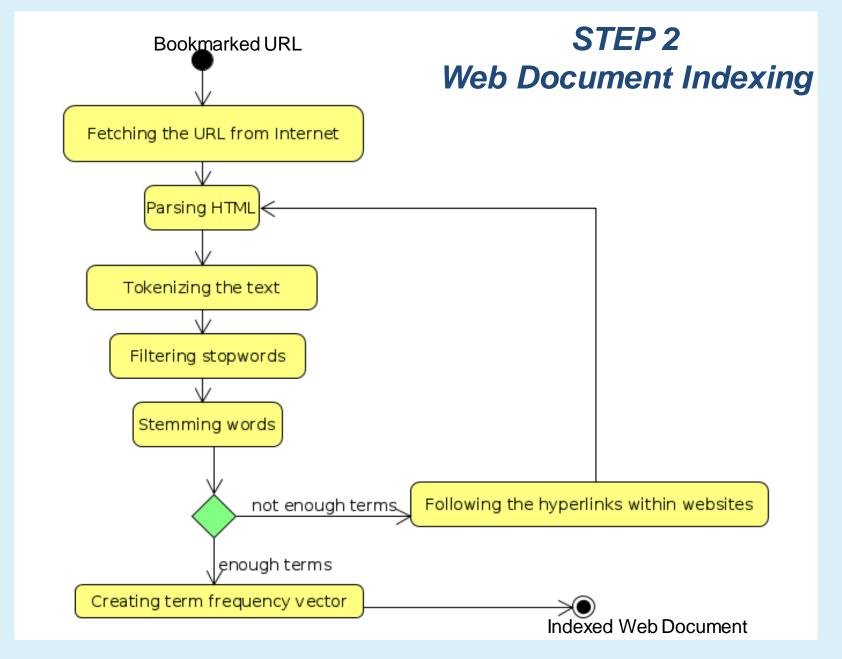


Fig. Activity diagram of a web document indexing

Step 3: CENTROID-BASED DOCUMENT CLASSIFICATION

- We used Rocchio's centroid algorithm, based on the vector space model
- Document is represented as vectors in a term space.
- For each category a centroid vector is created as a representative.
- Given k categories in the training set, this leads to k centroid vectors c1, c2, ck,

where each ci is the centroid for ith category, the category of new document vector is as follows:

$$\arg\max_{j=1,...,k}(\cos(\vec{x},\vec{c_j}))$$

CRITERIA FOR CATEGORISATION

- ☐ The classification is considered as unsuccessful if, score for tested document is not higher than the *THRESHOLD*.
- ☐ The tested document has to be similar to one of the training documents at least in 10%
- ☐ Else the classification will be to the other folder or system will ask for the creation of new folder.

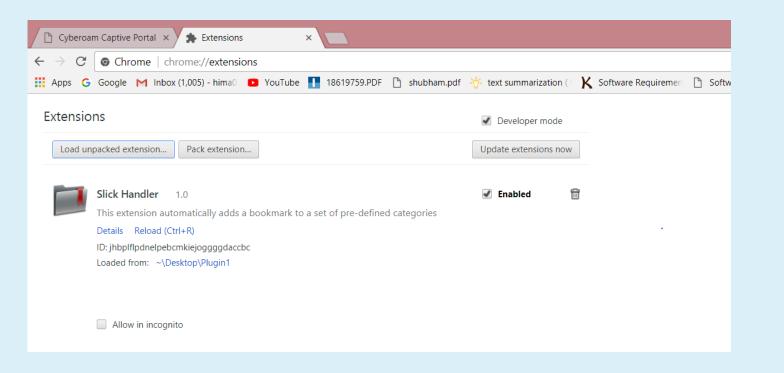
IMPLEMENTATION

```
C:\Users\H.P\Desktop\Plugin1\manifest.json - Notepad++
File Edit Search View Encoding Language Settings Tools Macro Run Plugins Window ?
 ] 🔒 🗎 🖺 🖺 🥫 🧸 | 🚜 🐚 🆍 | 🗩 🖒 | 🗷 🖒 🛍 😘 🖎 | 🤏 🕞 | 🚍 | 🚍 | 🚍 🖺 🖺 🖫 💹 🔑 🖆 💌 | 🗩 🗉 🕟
📙 popup js 🗵 📙 popup .html 🗵 📙 manifest json 🗵
          "manifest version": 2,
          "name": "Slick Handler",
   3
        "description": "This extension automatically adds a bookmark to a set of pre-defined categories",
   5
          "version": "1.0",
   6
      icons": {
            "48": "icon.png"
   8
   9
  10
          "permissions": [
  11
  12
          "http://*/*"
          "https://*/*
  13
  14
         1,
  15
      | "browser action": {
  16
            "default title": "Slick Handler",
  17
            "default icon": "icon.png",
  18
  19
            "default popup": "popup.html"
  20
  21
```

EXTENSION CODE: File 1: manifest.json

File 2: Popup.js

```
C:\Users\H.P\Desktop\Plugin1\popup.html - Notepad++
File Edit Search View Encoding Language Settings Tools Macro Run Plugins Window ?
🜏 🖴 🗎 🖺 🥦 🥱 🦓 🖊 🖎 🖍 🖍 🖍 🖒 🖒 🗩 🗢 🖒 🍇 🔍 🔍 🔍 🖂 🖫 🖺 🖺 🖫 🖫 🐼 🔊 🗀 📀 🗷 🗈 🕟
📙 popup js 🗵 📙 popup .html 🗵 📙 manifest json 🗵
      -<html>
  2
           <head>
  3
               <title>Slick Handler</title>
                                                                                                                           File 3:
               <script language='javascript' src='popup.js'></script>
  5
           </head>
                                                                                                                       Popup.html
  6
           <body>
               <form name='testForm'>
  8
                   <input type='button' id='alertButton' value='Add To Bookmarks'>
  9
                   <input style="margin-top: 16px" type='button' id='alertButton' value='View Bookmarks'>
 10
               </form>
           </body>
 12
       </html>
```

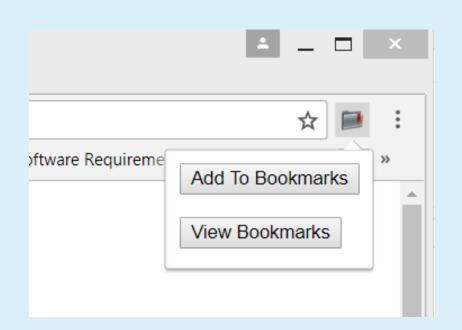


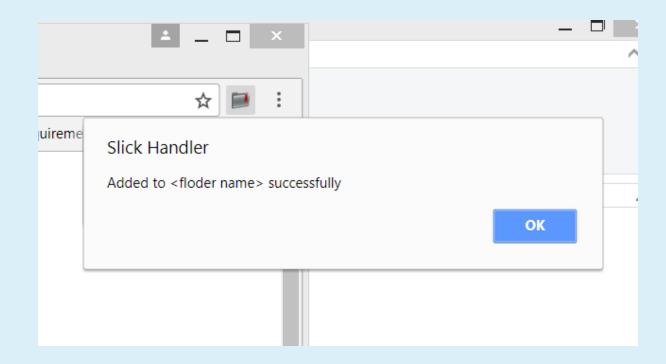
Browser Extension:

"Slick Handler"



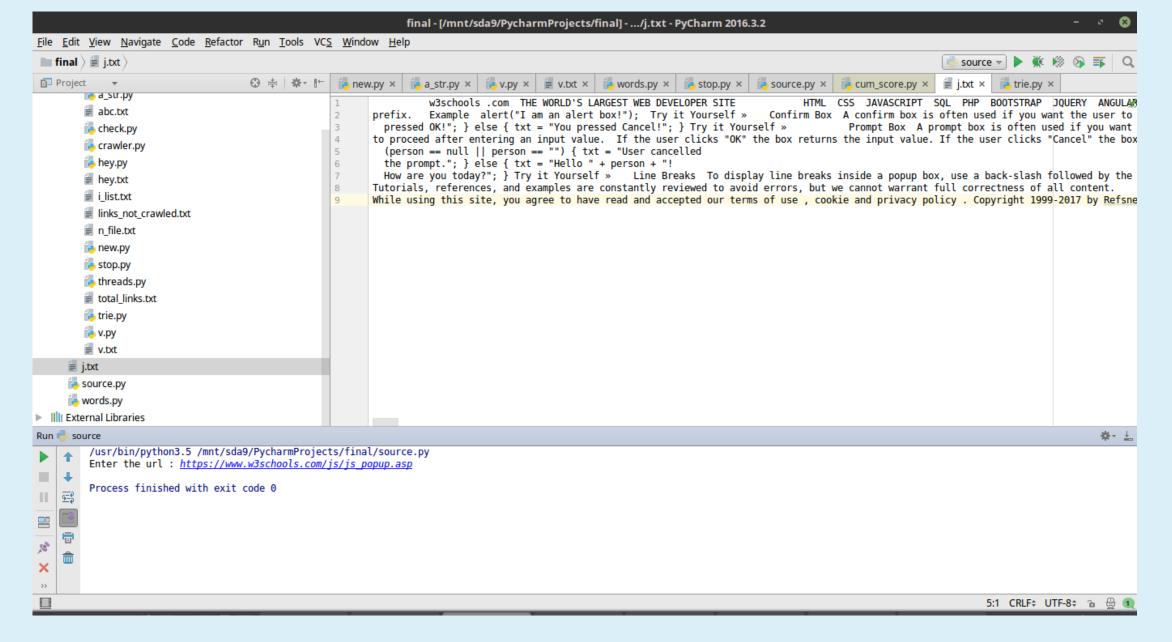
>>





```
from __future import division
from bs4 import BeautifulSoup
from bs4.element import Comment
'''from urllib.request import urlopen'''
import urllib
def tag visible (element):
    if element.parent.name in ['style', 'script', 'head', 'title', 'meta', '[document]']:
        return False
    if isinstance(element, Comment):
        return False
    return True
def text from html (body):
    soup = BeautifulSoup(body, 'html.parser')
    texts = soup.findAll(text=True)
    visible texts = filter(tag visible, texts)
    return u" ".join(t.strip() for t in visible texts)
url = input("Enter the url : ")
html = urlopen(url).read()
f = open('file1.txt', 'w')
text = text from html(html)
f.write(text)
f.close()
```

Fetching The URL: Web Crawler



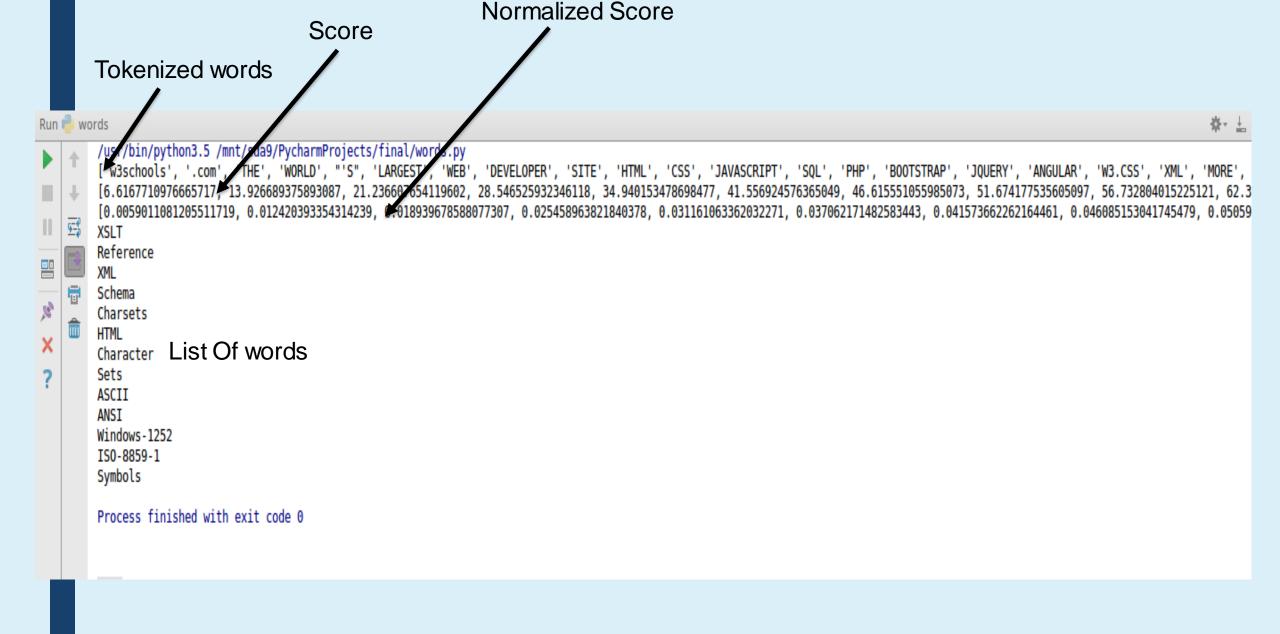
Text Extracted from a website through web crawler

```
from __future__ import division
from nltk.corpus import stopwords
from string import punctuation
from nltk.stem import PorterStemmer
from nltk import word_tokenize
from nltk import sent_tokenize
from operator import truediv
from sklearn.feature_extraction.text import TfidfVectorizer
import numpy as np
import sys
```

Python code for training and testing data

```
score=[]
for i in stem words:
    a=0
   if i in dict.keys():
            a=a+dict[i]
            score.append(a)
score1=[float(v) for v in score]
max=score1[0]
for i in score1:
   if i>max:
       max=i
for i in score1:
    score1[j]=i/max
   j=j+1
word dict={}
for i, j in zip(score1, word):
    word dict[i]=j
sort tot=sorted(score1)
final words=[]
for i in range (n-8, n):
    final words.append(word dict[sort tot[i]])
for i in final words:
   print(i)s
```

Python code for training and testing data



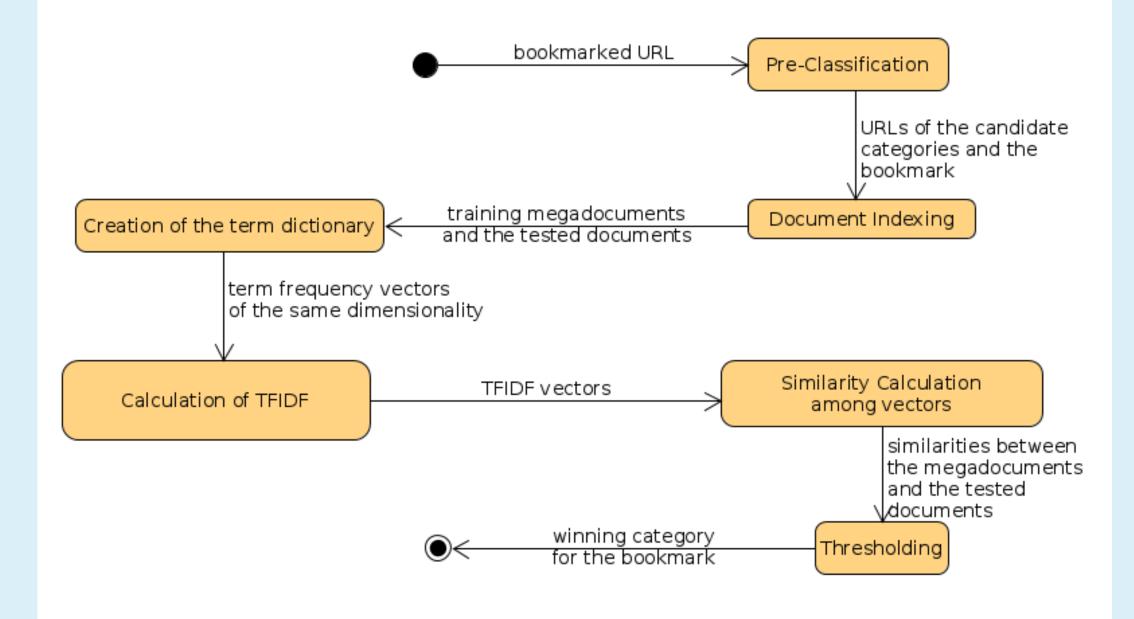


Fig. Overall System View

PROPOSED DELIVERABLES

- ☐ We proposed to deliver to a browser extension which will categorises the bookmarks
- ☐ The extension will also recommend the related web pages for the bookmarked sites
- The extension will also allow user to change the destination folder as per his requirement that is the extension will work on the automation as well as will work manually

SCOPE

□Our extension will be able to add a bookmark to the best fitting category from the pre-defined set of categories, which are

NEWS	BLOGS	ENTERTAINMENT	SONGS	EDUCATIONAL	SPORTS	OTHERS

- □ Users will be able to manage their bookmarks more easily and efficiently with the automation of bookmark manager
- □ For future aspects, we wish to add the search bar as well as wish to extend our domains further with the help of unsupervised machine learning