Big Data and Automated Content Analysis

Week 5 – Wednesday
Automated content analysis with NLP and
regular expressions

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Today

- ACA using regular expressions
 - Bottom-up vs. top-down What is a regexp?
 Using a regexp in Python
- 2 Natural Language Processing
 Stopword removal
 Stemming
 Parsing sentences
- 3 Take-home message & next steps

Automated content analysis using regular expressions

Bottom-up vs. top-down

Bottom-up

- Count most frequently occurring words (Week 2)
- Maybe better: Count combinations of words ⇒ Which words co-occur together? (Chapter 9, not obligatory)

We don't specify what to look for in advance

Bottom-up

- Count most frequently occurring words (Week 2)
- Maybe better: Count combinations of words ⇒ Which words co-occur together? (Chapter 9, not obligatory)

We don't specify what to look for in advance

Top-down

- Count frequencies of pre-defined words (like in BOW-sentiment analysis)
- Maybe better: patterns instead of words (regular expressions, today!)

We do specify what to look for in advance

Regular Expressions: What and why?

What is a regexp?

a very widespread way to describe patterns in strings

Regular Expressions: What and why?

What is a regexp?

- a very widespread way to describe patterns in strings
- Think of wildcards like * or operators like OR, AND or NOT in search strings: a regexp does the same, but is much more powerful

Regular Expressions: What and why?

What is a regexp?

- a very widespread way to describe patterns in strings
- Think of wildcards like * or operators like OR, AND or NOT in search strings: a regexp does the same, but is much more powerful
- You can use them in many editors (!), in the Terminal, in STATA ... and in Python

An example

From last week's task

- We wanted to remove everything but words from a tweet
- We did so by calling the .replace() method
- We could do this with a regular expression as well: [^a-zA-Z] would match anything that is not a letter

Basic regexp elements

Alternatives

Basic regexp elements

Alternatives

[TtFf] matches either T or t or F or f

Twitter|Facebook matches either Twitter or Facebook

. matches any character

Repetition

- * the expression before occurs 0 or more times
- + the expression before occurs 1 or more times

regexp quizz

Which words would be matched?

① [Pp]ython

regexp quizz

Which words would be matched?

- 1 [Pp]ython
- **2** [A−Z]+

regexp quizz

Which words would be matched?

- 1 [Pp]ython
- 2 [A-Z] +
- **3** RT ?:? @[a-zA-Z0-9]*

What else is possible?

If you google regexp or regular expression, you'll get a bunch of useful overviews. The wikipedia page is not too bad, either.

How to use regular expressions in Python

The module re

- re.findall("[Tt]witter|[Ff]acebook",testo) returns a list with all occurances of Twitter or Facebook in the string called testo
- re.findall("[0-9]+[a-zA-Z]+",testo) returns a list with all words that start with one or more numbers followed by one or more letters in the string called testo

The module re

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- re.findall("[0-9]+[a-zA-Z]+",testo) returns a list with all words that start with one or more numbers followed by one or more letters in the string called testo
- re.sub("[Tt]witter|[Ff]acebook","a social medium",testo)
 returns a string in which all all occurances of Twitter
 or Facebook are replaced by "a social medium"

How to use regular expressions in Python

The module re

```
re.match(" +([0-9]+) of ([0-9]+) points",line) returns

None unless it exactly matches the string line. If it

does, you can access the part between () with the

.group() method.
```

Example:

```
line=" 2 of 25 points"
result=re.match(" +([0-9]+) of ([0-9]+) points",line)
if result:
print ("Your points:",result.group(1))
print ("Maximum points:",result.group(2))
```

Your points: 2

Maximum points: 25

Possible applications

Data preprocessing

- Remove unwanted characters, words, ...
- Identify *meaningful* bits of text: usernames, headlines, where an article starts, . . .
- filter (distinguish relevant from irrelevant cases)

Possible applications

Data analysis: Automated coding

- Actors
- Brands
- links or other markers that follow a regular pattern
- Numbers (!)

Example 1: Counting actors

```
import re, csv
   from glob import glob
   count1_list=[]
    count2 list=[]
    filename list = glob("/home/damian/articles/*.txt")
5
6
    for fn in filename list:
7
8
       with open(fn) as fi:
         artikel = fi.read()
g
         artikel = artikel.replace('\n','')
10
11
         count1 = len(re.findall('Israel.*(minister|politician.*|[Aa]
12
              uthorit)', artikel))
         count2 = len(re.findall('[Pp]alest',artikel))
13
14
15
         count1_list.append(count1)
         count2_list.append(count2)
16
17
    output=zip(filename_list,count1_list, count2_list)
18
    with open("results.csv", mode='w',encoding="utf-8") as fo:
19
20
       writer = csv.writer(fo)
       writer.writerows(output)
21
```

Example 2: Which number has this Lexis Nexis article?

```
1
                                All Rights Reserved
2
                                 2 of 200 DOCUMENTS
3
5
                                    De Telegraaf
6
7
                               21 maart 2014 vrijdag
8
9
    Brussel bereikt akkoord aanpak probleembanken;
    ECB krijgt meer in melk te brokkelen
10
11
12
    SECTION: Finance: Blz. 24
    LENGTH: 660 woorden
13
14
             Europa heeft gisteren op de valreep een akkoord bereikt
15
    over een saneringsfonds voor banken. Daarmee staat de laatste
16
```

Example 2: Check the number of a lexis nexis article

```
All Rights Reserved
1
2
                                 2 of 200 DOCUMENTS
3
5
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6
7
                               21 maart 2014 vrijdag
8
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    Brussel bereikt akkoord aanpak probleembanken;
10
    ECB krijgt meer in melk te brokkelen
11
    SECTION: Finance; Blz. 24
12
    LENGTH: 660 woorden
13
14
    BRUSSEL Europa heeft gisteren op de valreep een akkoord bereikt
15
    over een saneringsfonds voor banken. Daarmee staat de laatste
16
    for line in tekst:
1
       matchObj=re.match(r" +([0-9]+) of ([0-9]+) DOCUMENTS",line)
       if matchObi:
           numberofarticle= int(matchObj.group(1))
4
           totalnumberofarticles= int(matchObj.group(2))
```

Practice yourself!

http://www.pyregex.com/

Natural Language Processing

NLP: What and why?

What can we do?

• remove stopwords

NLP: What and why?

What can we do?

- remove stopwords
- stemming

NLP: What and why?

What can we do?

- remove stopwords
- stemming
- parse sentences (advanced)

Natural Language Processing: **Stopword removal**

Natural Language Processing: **Stopword removal**

Have a look back at last week! The logic of the algorithm is very much related to the one of our first simple sentiment analysis!

Stopword removal: What and why?

Why remove stopwords?

- If we want to identify key terms (e.g., by means of a word count), we are not interested in them
- If we want to calculate document similarity, it might be inflated
- If we want to make a word co-occurance graph, irrelevant information will dominate the picture

Stopword removal: How

```
testo='He gives her a beer and a cigarette.'
2
   testonnovo=""
   mystopwords=['and','the','a','or','he','she','him','her']
   for verbo in testo.split():
       if verbo not in mystopwords:
          testonuovo=testonuovo+verbo+" "
6
```

What do we get if we do:

```
print (testonuovo)
```

Can you explain the algorithm?

We get:

```
>>> print (testonuovo)
'He gives beer cigarette. '
```

```
Why is "He" still in there?
How can we fix this?
```

Stopword removal

```
testo='He gives her a beer and a cigarette.'
   testonuovo=""
2
   mystopwords=['and','the','a','or','he','she','him','her']
   for verbo in testo.split():
5
       if verbo.lower() not in mystopwords:
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```

With *list comprehension* and the .join() method, you can achieve the same thing in one line:

```
tn2 = " ".join([w for w in testo.split() if w not in mystopwords])
```

Stopword removal

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testo='He gives her a beer and a cigarette.'
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With *list comprehension* and the .join() method, you can achieve the same thing in one line:

```
tn2 = " ".join([w for w in testo.split() if w not in mystopwords])
```

This is more efficient and more "pythonic", but may be more difficult to debug (especially if it gets more complicated)

Why do stemming?

- Because we do not want to distinguish between smoke, smoked, smoking, . . .
- Typical preprocessing step (like stopword removal)

Stemming

Stemming

(with NLTK, see Bird, S., Loper, E., & Klein, E. (2009). Natural language processing with Python. Sebastopol, CA: O'Reilly.)

```
from nltk.stem.snowball import SnowballStemmer
1
   stemmer=SnowballStemmer("english")
   frase="I am running while generously greeting my neighbors"
   frasenuevo=""
   for palabra in frase.split():
      frasenuevo=frasenuevo + stemmer.stem(palabra) + " "
```

If we now did print(frasenuevo), it would return:

```
i am run while generous greet my neighbor
```

Stemming and stopword removal - let's combine them!

```
from nltk.stem.snowball import SnowballStemmer
   from nltk.corpus import stopwords
   stemmer=SnowballStemmer("english")
   mystopwords = stopwords.words("english")
   frase="I am running while generously greeting my neighbors"
   frasenuevo=""
   for palabra in frase.lower().split():
       if palabra not in mystopwords:
8
          frasenuevo=frasenuevo + stemmer.stem(palabra) + " "
```

Now, print(frasenuevo) returns:

run generous greet neighbor

Perfect!

Stemming

Stemming and stopword removal - let's combine them!

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```

Now, print(frasenuevo) returns:

run generous greet neighbor

Perfect!

Or

Stemming

```
print(" ".join([stemmer.stem(p) for p in frase.lower().split() if p not
     in mystopwords]))
```

Stemming and stopword removal - let's combine them!

```
from nltk.stem.snowball import SnowballStemmer
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   from nltk.corpus import stopwords
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```

Now, print(frasenuevo) returns:

run generous greet neighbor

Perfect!

Or

```
print(" ".join([stemmer.stem(p) for p in frase.lower().split() if p not
     in mystopwords]))
```

In order to use nltk.corpus.stopwords, you have to download that module once. You can do so by typing the following in the Python console and selecting the appropriate package from the menu that pops up: import nltk nltk.download()

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ython 3.4)

Collections Corpora Models All Packages			
Identifier	Name	Size	Status 🏻
senseval sentiwordnet shakespeare sinica_treebank smultron state_union	SENSEVAL 2 Corpus: Sense Tagged Text SentiWordNet Shakespeare XML Corpus Sample Sinica Treebank Corpus Sample SMULTRON Corpus Sample C-Span State of the Union Address Corpus	2.1 MB 4.5 MB 464.3 KB 878.2 KB 162.3 KB 789.8 KB	not instal not instal not instal not instal not instal not instal
stopwords swadesh switchboard timit toolbox treebank udhr udhr2 unicode_samples universal_treebank	Stopwords Corpus Swadesh Wordlists Switchboard Corpus Sample TIMIT Corpus Sample Toolbox Sample Files Penn Treebank Sample Universal Declaration of Human Rights Corpu Universal Declaration of Human Rights Corpu Unicode Samples	8,5 KB 22,3 KB 772,6 KB 21,2 MB 244,7 KB 1,6 MB 1,1 MB	not instal not instal

Server Index: http://nltk.github.com/nltk_data/

Download Directory: /home/damian/nltk_data

In [5]: import nltk

In [6]: nltk.download()

4 A A A E A A E A E E

Parsing sentences

NLP: What and why?

Why parse sentences?

- To find out what grammatical function words have
- and to get closer to the meaning.

Parsing a sentence

```
import nltk
sentence = "At eight o'clock on Thursday morning, Arthur didn't feel
    very good."
tokens = nltk.word_tokenize(sentence)
print (tokens)
```

nltk.word_tokenize(sentence) is similar to sentence.split(), but compare handling of punctuation and the didn't in the output:

```
['At', 'eight', "o'clock", 'on', 'Thursday', 'morning', 'Arthur', 'did',
     "n't", 'feel', 'very', 'good', '.']
```

Parsing a sentence

Now, as the next step, you can "tag" the tokenized sentence:

```
tagged = nltk.pos_tag(tokens)
print (tagged[0:6])
```

gives you the following:

```
1 [('At', 'IN'), ('eight', 'CD'), ("o'clock", 'JJ'), ('on', 'IN'),
2 ('Thursday', 'NNP'), ('morning', 'NN')]
```

Parsing a sentence

Now, as the next step, you can "tag" the tokenized sentence:

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tagged = nltk.pos_tag(tokens)
print (tagged[0:6])
```

gives you the following:

```
[('At', 'IN'), ('eight', 'CD'), ("o'clock", 'JJ'), ('on', 'IN'),
('Thursday', 'NNP'), ('morning', 'NN')]
```

And you could get the word type of "morning" with tagged[5][1]!

Parsing sentences

More NI P

Look at http://nltk.org

Parsing sentences

More NLP

Look at http://spacy.io

Example: Named Entity Recognition with spacy

Terminal:

```
sudo pip3 install spacy
2
    sudo python3 -m spacy download nl # or en, de, fr ....
```

Python:

```
import spacy
   nlp = spacy.load('nl')
   doc = nlp('De docent heet Damian, en hij geeft vandaag les. Daarnaast is
        hij een onderzoeker, net zoals Anne. Ze werken allebei op de UvA')
   for ent in doc.ents:
5
      print(ent.text,ent.label_)
```

returns:

- Damian MISC Anne PER

 - Uva LOC

Parsing sentences

More NLP

Look at http://nlp.stanford.edu

More NI P

```
Look at
https://www.clips.
uantwerpen.be/pattern
```

Take-home message Take-home exam Next meetings

Take-home messages

What you should be familiar with:

- Possible steps to preprocess the data
- Regular expressions
- Word counts
- Where to look for NIP tasks

Next meetings

Friday

Write your own ACA script!

TAKE HOME EXAM

Handed out after Friday's meeting Deadline: Tuesday, 23.59