

# Regular expressions<sup>1</sup>

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<sup>1</sup>With materials used from "Speech and Language Processing" (2nd ed.), D. Jurafsky and J. H. Martin. 2007.

# Introduction

- Course:
  - Victor Kitov
  - Anna Potapenko
  - Murat Apishev
- Lectures+seminars
  - python+scikit-learn+numpy+matplotlib+...
  - linguistic packages: NLTK, pymorphy2, gensim, ...

## Recommended materials

- Books:
  - *Speech and Language Processing* (3rd ed. draft), D. Jurafsky and J. H. Martin.
  - *Speech and Language Processing* (2nd ed.), D. Jurafsky and J. H. Martin. 2007.
- Video-lectures:
  - D. Jurafsky & C. Manning: *Natural Language Processing*.
- Resources:
  - [Resource catalog for NLP](#)

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# Overview of text mining tasks

- Clustering
  - identify news about the same event
  - identify books on similar subject
- Topic modelling: probabilistic co-clustering of documents and terms.
- Classification
  - classify news into different categories: politics, sports, arts, etc.
  - assign documents to authors
    - are two documents written by the same person?
  - assign documents to genres:
    - survey, scientific article, remark, textbook, etc.

## Overview of text mining tasks

- Coreference resolution - identify expressions in a text referring to the same person or thing
  - The **music** was so loud that **it** couldn't be enjoyed.
  - Despite **her** difficulty, **Wilma** came to understand the point.
  - **Carol** told **Bob** to attend the party. **They** arrived together.
  - **Some of our colleagues** are going to be supportive. **These kinds of people** will earn our gratitude.
- Named entity recognition
  - locate and classify named entities in text into pre-defined categories
    - people names, organizations, locations, times, quantities, monetary values, percentages
  - Jim bought 300 shares of Acme Corp. in 2006. ->
  - **[Jim]** (*Person*) bought 300 shares of **[Acme Corp.]** (*Organization*) in **[2006]** (*Time*).

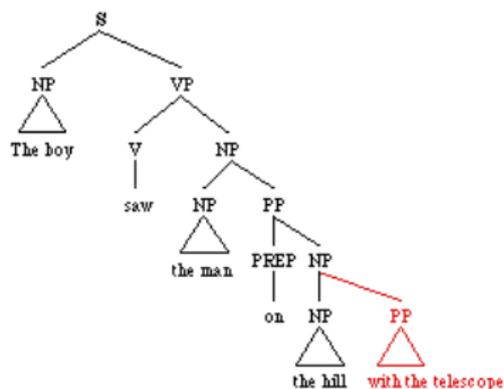
## Overview of text mining tasks

- Sentiment analysis (also known as opinion mining) - extract subjective attitudes from the text.
  - classify content into subjective (opinions) and objective (facts).
  - identify overall polarity
    - positive/negative or grade.
    - e.g.: negative movie review, rating 7 out of 10.
  - identify aspects-based attitude
    - extract individual aspects of entity
    - evaluate opinion about each aspect
    - e.g.: some cell phone review => design-excellent, battery-poor, ...
- Part-of-speech tagging

John saw the saw and decided to take it to the table.  
NNP VBD DT NN CC VBD TO VB PRP IN DT NN

# Overview of text mining tasks

- Syntactic parsing



- Automatic translation

- Je ne l'ai pas mangé depuis six jours ->
- I have not eaten it for six days.

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# Regular expressions

- re - python package for working with regular expressions.

## Simple match

RE	Example Patterns Matched
/woodchucks/	“interesting links to <u>woodchucks</u> and lemurs”
/a/	“ <u>M</u> ary Ann stopped by <u>M</u> ona’s”
/!/	“You’ve left the burglar behind again <u>!</u> ” said Nori

- Case sensitive: /Woodchucks/ will not match woodchucks

## Match any symbol from set

RE	Match	Example Patterns
/[wW]oodchuck/	Woodchuck or woodchuck	“ <u>W</u> oodchuck”
/[abc]/	‘a’, ‘b’, or ‘c’	“In uomini, in soldati”
/[1234567890]/	any digit	“plenty of <u>7</u> to 5”

# Regular expressions

- match any digit: `/[1234567890]/`
- match any uppercase letter:  
`/[ABCDEFGHIJKLMNOPQRSTUVWXYZ]/`

Shorter ways

RE	Match	Example Patterns Matched
<code>/[A-Z]/</code>	an upper case letter	“we should call it ‘ <u>D</u> renched Blossoms’ ”
<code>/[a-z]/</code>	a lower case letter	“ <u>m</u> y beans were impatient to be hoed!”
<code>/[0-9]/</code>	a single digit	“Chapter <u>1</u> : Down the Rabbit Hole”

- range: `/b-g/`
  - matches b, c, d, e, f, g.

Matching except set of characters

RE	Match (single characters)	Example Patterns Matched
<code>/[^A-Z]/</code>	not an upper case letter	“ <u>O</u> yfn pripetchik”
<code>/[^Ss]/</code>	neither ‘S’ nor ‘s’	“ <u>I</u> have no exquisite reason for’t”
<code>/[^\.]/</code>	not a period	“our resident Djinn”
<code>/[e^]/</code>	either ‘e’ or ‘^’	“look up <u>_</u> now”
<code>/a^b/</code>	the pattern ‘a^b’	“look up <u>a^</u> b now”

## Different number of occurrences

$p?$  matches pattern  $p$  or empty string.

None or single occurrence

RE	Match	Example Patterns Matched
<code>/woodchucks?/</code>	woodchuck or woodchucks	<u>“woodchuck”</u>
<code>/colou?r/</code>	color or colour	<u>“colour”</u>

- $p^*$  matches 0 or more occurrences of  $p$ :
  - `[], [p], [pp], [ppp], ...`
- $p^+$  matches 1 or more occurrences:
  - `[p], [pp], [ppp], ...`
- Recognizing sheep language: `baa!`, `baaa!`, `baaaa!`, ....
  - `/baaa*!/`
- `/cat|dog/` will match `[...cat...]` or `[...dog...]`.

# Anchors

**^** - start of string

- `/^The/` - will match «the» only at the start of the string
- [The red brown fox]

**\$** - end of string

- `/. *bushes$/` - will match «bushes» only at the end of the string
- [Fox jumped into the **bushes**.]

**\b** matches word boundary

- a word is sequence of letters, digits and underscore
- `/\bthe\b/` matches [in **the** trees]
- `/\bthe\b/` doesn't match [other].

## Other

- Special operators:

RE	Expansion	Match	Examples
\d	[0-9]	any digit	Party_of_5
\D	[^0-9]	any non-digit	Blue_moon
\w	[a-zA-Z0-9_]	any alphanumeric/underscore	Daiyu
\W	[^\w]	a non-alphanumeric	!!!!
\s	[\r\t\n\f]	whitespace (space, tab)	
\S	[^\s]	Non-whitespace	in_Concord

# Counts to match

<b>RE</b>	<b>Match</b>
*	zero or more occurrences of the previous char or expression
+	one or more occurrences of the previous char or expression
?	exactly zero or one occurrence of the previous char or expression
{ <i>n</i> }	<i>n</i> occurrences of the previous char or expression
{ <i>n</i> , <i>m</i> }	from <i>n</i> to <i>m</i> occurrences of the previous char or expression
{ <i>n</i> , }	at least <i>n</i> occurrences of the previous char or expression

## Matching reserved symbols

RE	Match	Example Patterns Matched
\*	an asterisk “*”	“K_A*P*L*A*N”
\.	a period “.”	“Dr. Livingston, I presume”
\?	a question mark	“Why don’t they come and lend a hand?”
\n	a newline	
\t	a tab	

# Substitutions

- /the (.\*)er they were, the \1er they will be/
  - will match «The bigger they were, the bigger they will be»
  - will NOT match «The bigger they were, the faster they will be»

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# Collocations

- Collocations are words that too frequently co-appear in text.
- Examples: New York, fast food, vice president, stock exchange, real estate, deja vu...

## Collocations extraction: t-test

- t-test for checking co-occurrence of  $w_i w_j$ :
  - define  $x = \mathbb{I}[w_i w_j]$
  - $\bar{x} = \frac{\#[w_i w_j]}{N}$ , where  $N$  is text length
  - test statistic:

$$\frac{\bar{x} - \mu}{\sqrt{s^2/N}} \rightarrow Student(N - 1) \rightarrow Normal(0, 1) \text{ for } N \rightarrow \infty$$

- where  $\mu = p(w_i)p(w_j) = \frac{\#[w_i]}{N} \frac{\#[w_j]}{N}$  - expected co-occurrence, given independence assumption.
- $s^2 = \bar{x}(1 - \bar{x})$  - sample variance.
- to be a collocation test statistic should be large.

## Collocations extraction: PMI

- Pointwise mutual information:

$$PMI(w_i w_j) = \frac{p(w_i w_j)}{p(w_i)p(w_j)}$$

Collocations extraction:  $\chi^2$  Person test

$\chi^2$  Pearson test for independence:

$$\begin{aligned}
 TS = & N \frac{[p(w_i w_j) - p(w_i)p(w_j)]^2}{p(w_i)p(w_j)} + N \frac{[p(w_i \bar{w}_j) - p(w_i)p(\bar{w}_j)]^2}{p(w_i)p(\bar{w}_j)} \\
 & + N \frac{[p(\bar{w}_i w_j) - p(\bar{w}_i)p(w_j)]^2}{p(\bar{w}_i)p(w_j)} + N \frac{[p(\bar{w}_i \bar{w}_j) - p(\bar{w}_i)p(\bar{w}_j)]^2}{p(\bar{w}_i)p(\bar{w}_j)}
 \end{aligned}$$

$$TS \approx N \frac{[p(w_i w_j) - p(w_i)p(w_j)]^2}{p(w_i)p(w_j)}$$

$$TS \sim \chi^2(1)$$

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# Tokenization

- Tokenization: division of text into tokens:
  - Foxes are small-to-medium-sized, omnivorous mammals. Foxes are slightly smaller than a medium-size domestic dog.
  - [Foxes] [are] [small]-[to]-[medium]-[sized], [omnivorous] [mammals]. [Foxes] [are] [slightly] [smaller] [than] [a] [medium]-[size] [domestic] [dog].
- Break dashed words? [medium]-[size] or [medium-size]?
- Count or not punctuation?
  - may reveal emotions - e.g. for sentiment analysis
  - useful for splitting into sentences , phrases=>text understanding.
  - useful for writer identification

# Utterance

- Count or not utterance?  
«*I do uh main- mainly business data processing*»
  - types of utterances:
    - fillers: uh, um, e-mmm
    - fragments: like [*main-*]
    - may reveal emotions - e.g. for sentiment analysis
      - have different meaning, like uh, um.
    - useful in text processing-utterance begin new clause, idea.
    - useful for speaker identification

## Stop words, capitalization

- Remove stop words?
  - and or not but,....
  - stop-words are corpus and task dependent
    - e.g. corpus of requests to city mayor - his name will be in all documents.
- Leave capitalization?
  - e.g. convert They->they?
  - capitalization is informative for, e.g., POS-tagging and named entity recognition.
  - for document classification, topic modelling - mostly not important.
  - may lose original meaning, like after US->us.

# Standardization

- Standardize words or not?
  - stemming
    - remove variable endings with fixed rules
  - lemmatization
    - replace wordform with lemma
    - using dictionary
    - we look for words with

## Stemming

- Most popular stemmer - Porter stemmer
- Stemmer as a cascade of deterministic rules, such as:

**ATIONAL** → **ATE** (e.g., relational →

**ING** →  $\epsilon$  if stem contains vowel

**SSES** → **SS** (e.g., grasses → gra

Still makes errors of:

- overgeneralization:
  - organization->organ
  - policy->police
- undergeneralization:
  - analysis->analyzes
  - European->Europe

# Segmentation

- We can segment not into words, but into larger strings
  - collect phrase statistics
  - detect authorship, plagiarize
- Segment into sequences of symbols
  - Man on a roof -> ma, an, n\_, \_o, on...
  - detect statistics of syllables, identifying language
    - e.g. classify recipes to countries of origin.
    - risotto, spaghetti,

# Sentence segmentation

- Sentence segmentation
  - natural unit of analysis for
    - POS tagging, syntactic analysis.
  - [!], [?] unambiguously identify sentence end
  - [.] - not necessarily:
    - Mr.Johnson travelled to central office of Microsoft Inc. in the U.S.A.
    - we need to build a classifier on segmented corpus.
    - using dictionary of abbreviations may help.

