

## Master of Technology in Knowledge Engineering

### Text Mining

# Linguistic and Knowledge Resources

**Fan Zhenzhen**  
**Institute of Systems Science**  
**National University of Singapore**  
email: zhenzhen@nus.edu.sg

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## Agenda

- Linguistic/knowledge resources and their roles in text mining
- Dictionaries
  - General dictionaries
  - Synonym dictionaries
  - WordNet
  - Sentiment/Opinion Lexicon
- Defining patterns using regular expressions



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## Objectives

- To be introduced to different types of resources
- To understand the roles of linguistic and knowledge resources
- To learn how to define such resources



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## Linguistic Resources

- Linguistic resources: sets of language data and descriptions in machine readable form
- Used for building text mining systems
  - Corpora - to provide examples for statistical methods to work
- Or for improving text mining systems, needed by various processing steps
  - **Dictionaries** - valid terms, POS information, list of stop words, or words to be filtered
  - **Terminologies** - special domain words and phrases
  - **Patterns/rules** - for information extraction



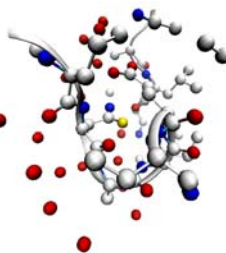
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## Knowledge Resources

- Taxonomy and ontology – a hierarchical conceptual model to map terms to concepts
- Prerequisite for advance TM, together with terminology lexicon
  - E.g. to derive complex information such as temporal, causal, conditional and other types of semantic relations between biomedical entities instead of simple associations
- More details in module

*“Advanced Topics in Text Mining”*



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## Dictionaries



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## Dictionaries

- Text analytics systems may be equipped with dictionaries in different languages for various purposes.
  - General domain dictionaries for more accurate tokenization, stemming, and POS tagging.
  - Terminology dictionaries for special domains or tasks, e.g.
    - Biomedical domain
    - Customer Relation Management
    - IT
    - Market Intelligence
    - Opinions Mining, etc.



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## Valid Term Dictionary

- A list of valid terms in the language in concern
- Or as dictionary for terms to be used in the term vector (e.g. R Text Mining package)
  - Only terms in the dictionary appear in the document term vector or matrix.
  - It helps to restrict the dimension of the matrix a priori and to focus on specific terms for distinct text mining contexts.
- It may include useful information such as POS



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## Filter Dictionary

- Also known as Stopword List / exclusion dictionary
- To support the stopwords removal step in preprocessing
- A list of very common words
  - usually functional words like *preposition, conjunction*, etc.
  - or words that are unimportant for the mining task
- Example stopwords list (not complete):

|                |               |                |                |
|----------------|---------------|----------------|----------------|
| <i>a</i>       | <i>an</i>     | <i>because</i> | <i>before</i>  |
| <i>about</i>   | <i>and</i>    | <i>been</i>    | <i>being</i>   |
| <i>above</i>   | <i>any</i>    | <i>before</i>  | <i>below</i>   |
| <i>after</i>   | <i>are</i>    | <i>being</i>   | <i>between</i> |
| <i>again</i>   | <i>aren't</i> | <i>below</i>   | <i>both</i>    |
| <i>against</i> | <i>as</i>     | <i>between</i> | <i>but</i>     |
| <i>all</i>     | <i>at</i>     | <i>both</i>    | <i>by</i>      |
| <i>am</i>      | <i>be</i>     | <i>Been</i>    | <i>...</i>     |

From <http://www.ranks.nl/resources/stopwords.html>



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## Synonym Dictionaries

- Also known as substitution dictionary
- to group similar words under one term
- Typically for known synonyms, user-defined synonyms

***dislike, detest***

- Also a direct way to deal with common misspellings with the correct spelling

***dislike, dilike***

- Can be used as a hard way to deal with inflections if no stemmer is used

***like, likes, liked***



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## Synonym Dictionaries

- Typically synonym words are listed in a file for string match
- Some tools/applications allow some flexibility in stating whether the synonyms should be matched
  - Strictly as it appears in the definition, disallowing inflected forms
  - With any word starting with the term
  - With any word ending with the term



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## WordNet

- A large lexical database of English
- Created and maintained by the Cognitive Science Laboratory of Princeton University
- Nouns, verbs, adjectives and adverbs are grouped into sets of cognitive synonyms (*synsets*), each expressing a distinct concept

### Number of words, synsets, and senses

| POS       | Unique Synsets |        | Total            |
|-----------|----------------|--------|------------------|
|           | Strings        |        | Word-Sense Pairs |
| Noun      | 117798         | 82115  | 146312           |
| Verb      | 11529          | 13767  | 25047            |
| Adjective | 21479          | 18156  | 30002            |
| Adverb    | 4481           | 3621   | 5580             |
| Totals    | 155287         | 117659 | 206941           |

Statistics from WordNet website  
<http://wordnet.princeton.edu/wordnet/man/wstats.7WN.html>



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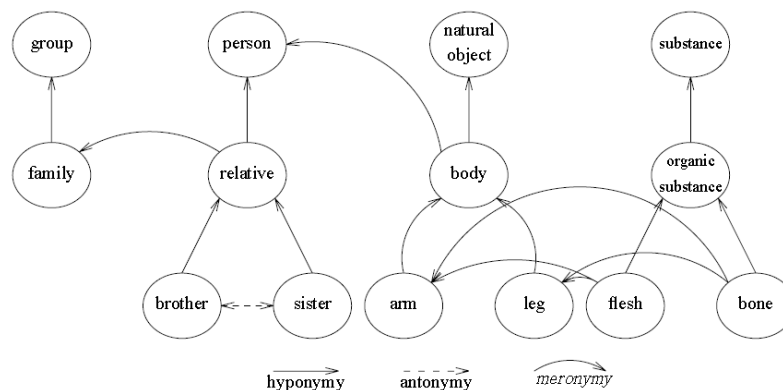
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## WordNet

- Synsets are linked by conceptual-semantic and lexical relations
  - Lexical relations
    - Synonymy – e.g. *shut* and *close*, *happy* and *joyful*
    - Antonymy – e.g. *wet* and *dry*, *young* and *old*, *happy* and *sad*
    - Morphological relations
  - Semantic relations
    - Hyponymy (or ISA relation, super-subordinate relation) – e.g. *apple* and *fruit*, *bed* and *furniture*, *communicate* and *talk* and *whisper*
    - Meronymy (part-whole relation) – e.g. *leg* and *chair*
  - And more...

## WordNet

Figure 2. Network representation of three semantic relations among an illustrative variety of lexical concepts



From *Nouns in WordNet: A Lexical Inheritance System*

## WordNet

- Example information in Wordnet for “happy”:

### Adjective

- (37) **S: (adj) happy#1** (enjoying or showing or marked by joy or pleasure)
- (2) **S: (adj) felicitous#2, happy#2** (marked by good fortune)
- **S: (adj) glad#2, happy#3** (eagerly disposed to act or to be of service)
- **S: (adj) happy#4, well-chosen#1** (well expressed and to the point)

- Expanded view:

- (37) **S: (adj) happy#1** (enjoying or showing or marked by joy or pleasure)
  - *see also*
  - *similar to*
    - **S: (adj) blessed#6** (characterized by happiness and good fortune)
    - **S: (adj) blissful#1** (completely happy and contented)
    - **S: (adj) bright#9** (characterized by happiness or gladness)
    - **S: (adj) golden#2, halcyon#2, prosperous#3** (marked by peace and prosperity)
    - **S: (adj) laughing#1, riant#1** (showing or feeling mirth or pleasure or happiness)
  - *attribute*
  - *antonym*
    - **W: (adj) unhappy#1** [Opposed to: **happy**] (experiencing or marked by or causing sadness or sorrow or discontent)



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## WordNet

- Free and open source
- Proved useful for a wide range of Natural Language Processing applications
  - Word sense disambiguation
  - Word semantic distance measuring
  - Mono- and cross-lingual Information retrieval,
  - Question-answering systems
  - Machine translation
  - Document structuring and categorisation



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## Sentiment/Opinion Lexicon

- Essential resources required for Opinion Mining to detect sentences containing subjective opinions.
- also known as *sentiment words*, *opinion words*, *polar words*, or *opinion-bearing words*.
- Lexicons or dictionaries of words or phrases that convey *positive* or *negative* sentiments, for example:

*beautiful, wonderful, amazing...*  
*bad, poor, awful...*

- Such sentiment/opinion lexicon can be manually compiled, which can be labor intensive and time consuming.



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## Sentiment/Opinion Lexicon



- Or they can be generated automatically from dictionaries
  - Start with a small set of manually collected seed sentiment words
  - Search in WordNet or other online dictionaries for their synonyms and antonyms
  - Add the newly found words to the seed list
  - Begin the next iteration until no more new words can be found
- Another approach is to generate from corpus
  - Start with some seed words and identify more sentiment words and their orientation using linguistic rule or conventions on connectives (*and*, *or*)
  - More complex



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## Challenges in Using Opinion Lexicon

- An opinion word's opinion orientation can be sensitive to its context.
  - E.g. *long* – **positive** or **negative**?
    - “The battery life is very *long*” 
    - “The queue at the counter is very *long*” 
- Sarcasm, in which the speakers say the opposite of what they mean
  - E.g. “What a *great* phone! It stopped working in two days.”



## Sentiment/Opinion Lexicon

- Some Sentiment Lexicons are publically available
  - General Inquirer lexicon:  
[http://www.wjh.harvard.edu/~inquirer/spreadsheet\\_guide.htm](http://www.wjh.harvard.edu/~inquirer/spreadsheet_guide.htm)
  - Sentiment lexicon (Liu Bing):  
<http://www.cs.uic.edu/~liub/FBS/sentiment-analysis.html#lexicon>
  - OpinionFinder subjectivity lexicon:  
<http://www.cs.pitt.edu/mpqa/lexicons.html>
  - SentiWordNet – assign each synset of WordNet sentiment scores  
<http://sentiwordnet.isti.cnr.it/>

## Defining Patterns using Regular Expressions



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### Defining patterns/rules

- With regular expression, we can extract strings containing certain characters, or not containing certain characters, or strings with pre-specified patterns of letters or numbers.
- Such patterns can be defined in a very compact way
  - E.g. regular expression for email addresses  
`[A-Z0-9._-]+@[A-Z0-9.-]+\.[A-Z]{2,4}`
  - Strings matching this expression can then be extracted
    - E.g. [zhenzhen@nus.edu.sg](mailto:zhenzhen@nus.edu.sg)

Regular expressions are very useful in extracting concepts expressed in a certain way, e.g. *currency*, *dates*, *e-mail addresses*, *phone numbers*, etc.



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## Common Operators

- Special characters (operators) are used to define character patterns

| Operator | Purpose  |
|----------|--|
| .        | (period) Match any single character<br>E.g. .in matches both <b>Windows</b> , and <b>Linux</b>                     |
| ^        | Match the empty string that occurs at the beginning of a line or string<br>E.g. ^tre will not match <b>stretch</b> |
| \$       | Match the empty string that occurs at the end of a line  |
| \d       | Match any single digit   |
| \D       | Match any single non-digit character   |
| \w       | Match any single alphanumeric character  |

## Common Operators

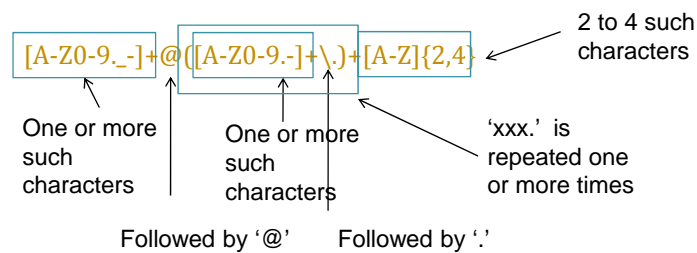
| Operator | Purpose   |
|----------|---|
| ?        | Match the preceding character 0 or 1 time<br>E.g. colou?r matches <b>color</b> (0) and <b>colour</b> (1)  |
| *        | Zero or more of the preceding character<br>E.g. tre* matches <b>tree</b> (2), <b>tread</b> (1), and <b>trough</b> (0)   |
| +        | Match the preceding character 1 or more times<br>E.g. tre+ matches <b>tree</b> , and <b>tread</b>   |
| [...]    | Match anything inside the square brackets for one character position once<br>E.g. [0-9] matches any character in the range 0-9<br>[abc] matches <b>a</b> , <b>b</b> , or <b>c</b> |
| [^...]   | Match any character excluding those in the square brackets<br>E.g. [^A-M]in matches <b>Windows</b> , but not <b>Linux</b>   |

## Common Operators

| Operator | Purpose   |
|----------|---|
| {n}      | Match the preceding character, or character range, n times<br>E.g. [0-9]{3}-[0-9]{4} matches local phone number like 123-4567                     |
| {n,m}    | Match the preceding character at least n times but not more than m times<br>E.g. [A-Z]{2,4} matches <i>com</i> , <i>sg</i> , but not <i>abcde</i> |
| ()       | Group parts of search expression together   |
|          | Separate two alternative values<br>E.g. gr(a e)y matches both <i>gray</i> and <i>grey</i>   |
| \b       | Match empty string, frequently used to indicate a word boundary<br>E.g. \bhis\b matches <i>his</i> only, not <i>this</i> or <i>history</i>        |

## Regular Expression

- Take a look at our email pattern regex again:



## Reference and Resources

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