Practical Text Analytics: Introduction

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Info

- Github repo for tutorials and coding materials
  `https://github.com/fanstats/text-analytics`

- References

  📄 Murugan Anandarajan, Chelsey Hill, Thomas Nolan
  Practical text analytics: maximizing the value of text data
  Springer International Publishing, 2019
  `https://link-springer-com.proxy.lib.iastate.edu/`
  `book/10.1007%2F978-3-319-95663-3`

- Practical skills with R
  Prepare a corpus of documents, convert them to a data matrix and
  illustrate statistical methods available for different analysis.

  Packages: `tm, tidytext, tidyr, tidyverse, hunspell, etc.`

Introduction: Text data

- Text data
  - Product reviews, social media posts, customer feedback, etc.

* Concepts & topics
* Sentiment & Emotions
* Behaviors

Source: **https://www.wordstream.com/social-media-marketing**

Introduction: Text data

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  - **Unstructured data**
    * Structured data: stored and organized in typical databases like Excel, Google Sheets, and SQL.
    * Unstructured data: everything else..

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Traveller Id	Traveller star	location	cleanliness	Service	Sleep Quality	Value	Rooms	Trip Type	Text reviews
Filippo S	5		5	5	5			business	I had the most incredible experience in this hotel. I
Explorer078	3	1	3	2	2	3	3	business	Getting in and out of hotel is extremely hazardous
hazimismail	4	3	3	4	4	3	4	business	Wverything was fine with the hotel. Rooms were a
Barranquita	5		5	5	5	5	5	family	Free airport shuttle on time both at arrival and dep
Josephine B	5								HGI Miami APW was a very nice place to unwind af Cruise. They were very accommodating (able to che free water bottles). It is not too far from the airpor free shuttle) and has many fast-food restaurants (N Starbucks, Dunkin Donuts) within walking distance. Also appreciate the onsite washing machines and c a small fee) and morning breakfast (another fee). Highly recommend.
Isabelle W	5	5	5	5	5	5	5	couple	Just check-out of my favorite Miami hotel - again, th
mikeslady	4								Great breakfast buffet, nice seating area in lobby fo
Frankie_Sur	4	3	4	4	5	4	5	family	Very Nice Hotel , Clean rooms, free shuttle to Airpo

Figure: Customer reviews for Hilton Garden Inn-Miami Airport (Dec-Jan, 2019)

Source: **https://www.tripadvisor.com/**

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  - **Big data**
    * Massive volumes
    * Many different file types
    * High-speed data creation

Introduction: Text analytics

- Text analytics

    - Uses: including but not limited to email filtering, product suggestions, fraud detection, opinion mining, trend analysis,search engines, etc.

    - Goals and objectives: understanding semantic information, text summarization, classification, clustering, etc.

    - Information retrieval and extraction

    - Data mining

        * Quantitative results

        * Graphs, tables and other sorts of visual reports.

Figure: Word cloud of text analytics article titles and abstracts

Source: **Practical text analytics: maximizing the value of text data**

Introduction: Text analytics

Figure: Word cloud of text analytics job titles

Source: **Practical text analytics: maximizing the value of text data**

Figure: Word cloud of the skills required for text analytics jobs

Source: **Practical text analytics: maximizing the value of text data**

Introduction: Text analytics process

Text analytics process

- Planning the text analytics projects

- Text preparing and preprocessing

- Unsupervised and supervised analysis
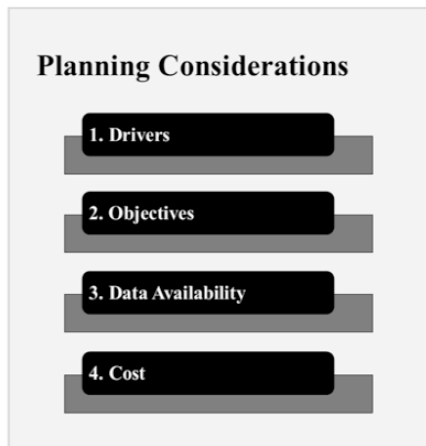
- Result interpretations

Introduction: Text analytics process

Planning the text analytics projects

- Initial consideration

- Problem framing

- Data generation

- Method and implementation selection

- Inference and decision-making

Initial considerations

Figure: Initial planning considerations

Source: **Practical text analytics: maximizing the value of text data**

Introduction: Text analytics process

Problem framing

- Identify analysis problems

- Decide types of inference

  - **Deductive inference**

    - Research questions $\Rightarrow$ Theory $\Rightarrow$ Hypothese $\Rightarrow$ Conceptualization

    - Theory-driven coding schemes

Introduction: Text analytics process

Problem framing

- Identify analysis problems

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    - **Deductive inference**
        - Research questions $\Rightarrow$ Theory $\Rightarrow$ Hypothese $\Rightarrow$ Conceptualization
        - Theory-driven coding schemes
    - **Inductive inference**
        - Data-driven coding schemes
        - Create a generalizable theory

Introduction: Text analytics process

Data generation

- Define research scope and purpose
  - Unitization
    - Sampling unit: each row of data or text document
    - Recording unit: piece of text that is categorized in the coding step
    - Context unit: define the amount of text that can be used

Introduction: Text analytics process

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- Collect text data
  - Sampling techniques
    - Non-probability sampling: convenience sampling, relevance sampling
    - Probability sampling: systematic sampling, stratified sampling

Introduction: Text analytics process

- Data quality

Figure: Quality dimensions of text data

Example: Business policies on human rights (Preuss and Brown, 2012)

- Inductive inference
- Research questions: frequency and content of human rights policies in large corporations
- Relevance sampling: all companies on the Financial Times Stock Exchange 100 (FTSE 100) list
- Sampling unit: each firm's website
- Context unit: corporate codes of conduct ans supporting documents
- Recording unit:
  whether a company had a human rights policy: corporate website

Introduction: Text analytics process

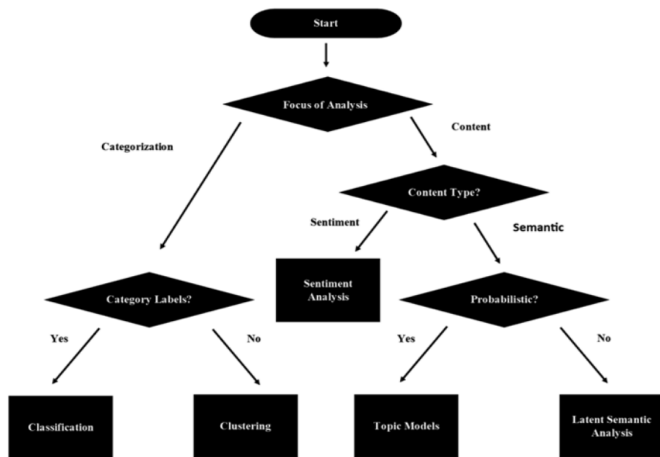Different methods based on our objective(s)

Figure: Flow chart of what-to-do based on our objective

Source: **Practical text analytics: maximizing the value of text data**

Different methods based on our objective(s):(contd.)

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- If our objective is not to categorize or classify the documents, rather we're concerned with the content of the documents, we should go for the other type:

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    - To identify semantic (*"linguistic and philosophical study of meaning in language..."* - Wikipedia) information in the documents, there are two classical analysis methods available: **latent semantic analysis (LSA)** and **topic models**. Topic models are probabilistic, generative models; while LSA deals with dimension reduction methods.

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- Other methods, like text summarization also exits.