

1 Text Preprocessing

1.1 Tokenization

- **Task:** Split arbitrary input text into (linguistically salient) *tokens*.
- **Motivation:** Dealing with unrestricted input texts in a holistic manner is unfeasible.
- **Method:** Traditionally, whitespace and punctuation have been used to identify token boundaries.
- **Problems:**
 - Numbers (“42.24”, “1,001”),
 - Times (“4:20”, “15:00”)
 - Abbreviations (“M.I.T.”, “Ph.D.”)
 - Collocations & Idioms (“New Haven”, “kind of”)
- **Workaround:** 2-stage analysis
 - *Stage 1:* Identify *broad segmentation units (BSUs)*: split input on all punctuation characters and whitespace subsequences.
 - *Stage 2:* Identify *final segmentation units (FSUs)*: map BSU subsequences to FSUs.
- **Example**
 - *Input:*
1,000 Ph.D.s will meet in New Haven.
 - *Stage-1 Output:*
(1)(,)(000)(Ph)(.)(D)(.)(s)(will)(meet)(in)(New)(Haven)(.)
 - *Stage-2 Output:*
(one)(thousand)(P)(H)(Ds)(will)(meet)(in)(New Haven)(.)

1.2 Expansions

- **Task:** Expand numbers, abbreviations, and acronyms to canonical orthographic representations – typically performed as part of the *2nd* stage of tokenization.
- **Motivation:** Allows uniform treatment of input data by later stages.
- **Methods:**
 - Full-form token subsequence rewrite grammar.
 - Finite-state transduction.
- **Problems:**

- *Numbers*
 - * Full-form treatment not possible for infinite set.
 - * Differing conventions for pronunciation of cardinals, ordinals, times, and years.
- *Abbreviations*
 - * 1 : n mapping from abbreviations to canonical orthographic forms, e.g.
 - German “t_gl.” → {*t_gglich, t_gglich_er, t_gglich_es, . . .*}
 - English “St.” → {*street, saint*}
- *Acronyms*
 - * Full-form treatment not practical.
 - * Some acronyms are spelled out (“CPU”, “BA”), while others are spoken as single words (“RAM”, “SCSI”).
- **Workaround(s):**
 - Probabilistic determination of “best” expansion (error-prone).
 - Context-dependent expansion heuristics (pre-empts “real” contextual analysis).
 - Additional markup for problematic tokens to be treated at a later stage.

1.3 Sentence Boundary Detection

- **Task:** Identify and mark sentence boundaries in input text.
- **Motivation:** Sentence boundaries (and types) influence prosodic parameters.
- **Method:**
 - *Sentence-terminal punctuation:* Traditionally, the punctuation characters “.”, “?”, “!”, and “:” have been used as sentence end markers.
 - *Sentence-initial capitalization:* For English, capitalization is often used as a cue for sentence boundary detection.
- **Problems:**
 - Token-internal punctuation (numbers, times, abbreviations, *etc.*)
 - Language-specific capitalization conventions
- **Workaround(s):**
 - Good tokenization and expansion modules can help reduce punctuation misinterpretation.
 - Probabilistic method described in Liberman and Church (1992).

1.4 Collocations

- **Task:** Identify and label collocations and idioms in the input token stream.
- **Motivation:** Prosody for collocations often does not conform to the “normal” rules their surface forms.
- **Method(s):**
 - *Condensation:* Treat like numbers and acronyms, “expansion” becomes “condensation”.
 - *Procrastination:* Ignore in preprocessing stage, analyze later (using output from morphology / tagger / chunker / parser).
- **Problems:**
 - *Condensation:* For some languages (German, French), collocational surface forms depend on morphosyntactic context (inflection), which makes them hard to detect early on.
 - *Procrastination:* Correct identification of collocations can provide crucial information for later stages of contextual analysis.

2 Morphological Analysis

- **Task:** Segment input tokens into *morph*¹ sequences.
- **Motivation:**
 - Pronunciation dictionary minimization.
 - Identification of morph boundaries can improve letter-to-sound transduction accuracy:
 - * English “bo**T**Her” vs. “ho**T**/**H**ouse”,
 - * German “Neben/**S**trasse” vs. “Demon**S**tra/tion”
 - Identification of inflectional morphology can restrict search space for later contextual analysis stages (PoS tagging), thereby improving efficiency.
 - Root/affix differentiation improves stress assignment accuracy.
- **Methods:**
 - *Full-form lexicon:* Impractical, inaccurate, and costly, but fast – useful for closed-class items.
 - *Procedural rules:* Must be painstakingly hand-crafted, impractical for heavily inflected languages.
 - *Declarative stem/affix association lexicon:* Minimal storage, but difficult to construct and inefficient to access at runtime.

¹A morph is either a root or an affix

- *Two-level morphology*: Expressable as a finite-state transducer (FST), quite efficient, and can even handle compounding.

- **Problems & Workarounds:**

- *Efficiency*: Many morphological analysis strategies require a good deal of processing power in their raw forms; workarounds usually involve pre-compiled indices used to speed lookup operations.
- *Indexing*: Indexing of roots and affixes for efficient access can result in large memory requirements (also applies FST methods); workarounds generally result in greater time complexity.
- *Robustness*: Misspellings and previously unknown tokens are not handled by any of the above methods in their strict forms; a workaround known as an *open lexicon strategy* allows unknown stems in analyses and reduces memory requirements, but can also harm accuracy.

References

- J. Allen, S. Hunnicut, and D. Klatt. *From Text to Speech: the MITalk system*. Cambridge University Press, 1987.
- T. Dutoit. *An Introduction to Text-to-Speech Synthesis*. Kluwer, Dordrecht, 1997.
- M. J. Liberman and K. W. Church. Text analysis and word pronunciation in text-to-speech synthesis. In S. Furui and M. M. Sondhi, editors, *Advances in Speech Signal Processing*. Dekker, New York, 1992.