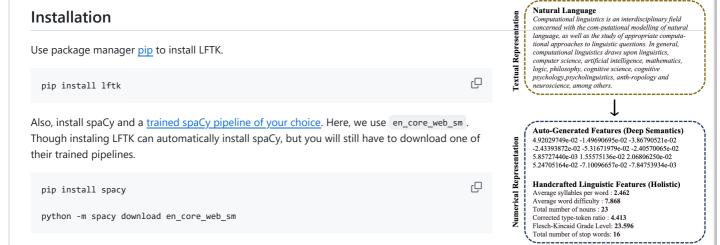


Much faster than LFTK's predecessor, <u>LingFeat</u>. This time is reported excluding spaCy processing time, which is not contained in LFTK.

https://github.com/brucewlee/lftk 1/5

• On More with SpaCy: LFTK is built on top of a popular NLP library named spaCy. Explore spaCy's pre-trained pipelines and get the most out of spaCy.

LFTK can calculate readability score, evaluate word difficulty, count number of nouns, and many more. There is much to explore in this package. Use our handcrafted features to support linguistic studies or build machine learning models.



Citation

```
ſĠ
@inproceedings{lee-lee-2023-lftk,
    title = "{LFTK}: Handcrafted Features in Computational Linguistics",
    author = "Lee, Bruce W. and Lee, Jason",
    booktitle = "Proceedings of the 18th Workshop on Innovative Use of NLP for Building Educational Applications (BEA 2023)",
    month = jul,
   year = "2023",
    address = "Toronto, Canada",
    publisher = "Association for Computational Linguistics",
    url = "https://aclanthology.org/2023.bea-1.1",
    pages = "1--19",
    abstract = "Past research has identified a rich set of handcrafted linguistic features that can potentially assist
various tasks. However, their extensive number makes it difficult to effectively select and utilize existing handcrafted
features. Coupled with the problem of inconsistent implementation across research works, there has been no categorization
scheme or generally-accepted feature names. This creates unwanted confusion. Also, no actively-maintained open-source library
extracts a wide variety of handcrafted features. The current handcrafted feature extraction practices have several
inefficiencies, and a researcher often has to build such an extraction system from the ground up. We collect and categorize
more than 220 popular handcrafted features grounded on past literature. Then, we conduct a correlation analysis study on
several task-specific datasets and report the potential use cases of each feature. Lastly, we devise a multilingual
handcrafted linguistic feature extraction system in a systematically expandable manner. We open-source our system to give the
community a rich set of pre-implemented handcrafted features.",
```

News

- We have the first round of use cases. Thank you for using LFTK!
 - o ICMI 2023 Gaze-Driven Sentence Simplification for Language Learners: Enhancing Comprehension and Readability
 - o Preprint Benchmarking and Explaining Large Language Model-based Code Generation: A Causality-Centric Approach
 - SLaTE 2023 Effective Neural Modeling Leveraging Readability Features for Automated Essay Scoring
 - Preprint Large Language Models and Financial Market Sentiment
- Come join our presentation at BEA @ ACL 2023.
- Preprint available on ArXiv! Here
- v.1.0.9 -> Documentation update! Keep track of our progress.
- v.1.0.8 -> 7 features that extracts conjunctions are deleted. These features are replaced by those extractin subordinating conjunctions and coordinating conjunctions.

Usage

```
import spacy
import lftk

# load a trained pipeline of your choice from spacy
# remember we already downloaed "en_core_web_sm" pipeline above?
nlp = spacy.load("en_core_web_sm")
```

```
# create a spaCy doc object
doc = nlp("I love research but my professor is strange.")

# initiate LFTK extractor by passing in doc
# you can pass in a list of multiple docs
LFTK = lftk.Extractor(docs = doc)

# optionally, you can customize how LFTK extractor calculates handcrafted linguistic features
# for example, include stop word? include puncutaion? maximum decimal digits?
LFTK.customize(stop_words=True, punctuations=False, round_decimal=3)

# now, extract the handcrafted linguistic features that you need
# refer to them as feature keys
extracted_features = LFTK.extract(features = ["a_word_ps", "a_kup_pw", "n_noun"])

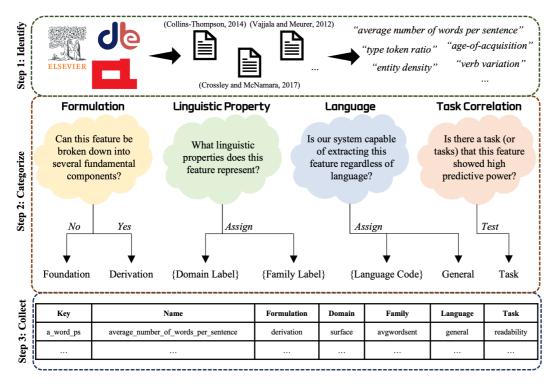
# {'a_word_ps': 8.0, 'a_kup_pw': 5.754, 'n_noun': 2}
print(extracted_features)
```

Also, read Essential Tips and To-Do Guides.

Deep Dive: Handcrafted Linguistic Features

TL;DR: Google Sheet of All Handcrafted Linguistic Features

Each handcrafted linguistic feature represents a certain linguistic property. We categorize all features into the broad linguistic branches of lexico-semantics, syntax, discourse, and surface. The surface branch can also hold features that do not belong to any specific linguistic branch. Apart from linguistic branches, handcrafted features are also categorized into linguistic families. The linguistic families are meant to group features into smaller subcategories, enabling users to search more effectively for the feature they need. All family names are unique, and each family belongs to a specific formulation. This means that the features in a family are either all foundation or all derivation. A linguistic family also serves as an important building block of our feature extraction system. LFTK as a program is essentially a linked collection of several feature extraction modules where each module represents a linguistic family.



Each handcrafted linguistic feature can either foundation or derivation. Derivation-type linguistic features are derived on top of foundation-type linguistic features. For example, the *total number of words* and the *total number of sentences* in a given text is a foundation feature. On the other hand, the *average number of words per sentence* is a derivation feature as it builds on top of the two aforementioned foundation features.

Each handcrafted linguistic feature also has an assigned language value. If the linguistic feature is universally applicable across languages, it is denoted "general". These general linguistic features can be used with any language given that spaCy has a supporting pipeline for that functionality in that language. This can be easily checked on spaCy pipelines. If the feature is designed for a specific language, like English, it is denoted with the specific language code.

Programmatically Searching Handcrafted Features

```
import lftk
# returns all available features as a list of dictionaries by default
searched_features = lftk.search_features()
# [{'key': 't_word', 'name': 'total_number_of_words', 'formulation': 'foundation', 'domain': 'surface', 'family': 'wordsent'},
print(searched_features)
# specify attributes
searched features = lftk.search features(domain = "surface", family = "avgwordsent")
# [{'key': 'a_word_ps', 'name': 'average_number_of_words_per_sentence', 'formulation': 'derivation', 'domain': 'surface', 'fami.
print(searched_features)
# return pandas dataframe instead of list of dictionaries
searched_features = lftk.search_features(domain = 'surface', family = "avgwordsent", return_format = "pandas")
# key
                                                  name formulation domain
                                                                                 family
#4 a word ps
                average_number_of_words_per_sentence derivation surface avgwordsent
#5 a_char_ps average_number_of_characters_per_sentence derivation surface avgwordsent
#6 a_char_pw average_number_of_characters_per_word derivation surface avgwordsent
print(searched_features)
```

Attribute: domain

- surface : surface-level features that often do not represent a specific linguistic property
- lexico-semantics: attributes associated with words
- discourse: high-level dependencies between words and sentences
- syntax: arrangement of words and phrases

Attribute: family

- wordsent: basic counts of words and sentences
- worddiff: difficulty, familiarity, frequency of words
- partofspeech: features that deals with part of speech properties, we follow the universal POS tagging scheme
- entity: named entities or entities such as location or person
- avgwordsent : averaging wordsent features over certain spans
- avgworddiff: averaging worddiff features over certain spans
- avgpartofspeech: averaging partofspeech features over certain spans
- avgentity : averaging entity features over certain spans
- lexicalvariation: features that measure lexical variation (that are not TTR)
- typetokenratio: type token ratio is known to capture lexical richness of a text
- readformula: traditional readability formulas that calculate text readability
- readtimeformula: basic reading time formulas (in seconds)

Attribute: language

- general: LFTK can extract this feature in a language-agnostic manner when supplied with an appropriate spaCy pipeline
- en: LFTK can extract this feature in English only

Essential Tips and To-Do Guides

Q: How to extract features by group? Do I have to specify each feature individually?

No. We have a good way around, using the convenient search function. First, think about how you want to search for your handcrafted linguistic features. In this case, we only want wordsent family features that generally work across languages.

```
import lftk

# specify attributes and (IMPORTANT) set return_format to "list_key"
searched_features = lftk.search_features(family = "wordsent", language = "general", return_format = "list_key")

#['t_word', 't_stopword', 't_punct', 't_uword', 't_sent', 't_char']
print(searched_features)
```

How is this possible? search_features function returns all available features by default and a user can restrict the returned features by specifying attributes. This is analogous to asking the function to "return all features that are {attribute 1}, {attribute 2}, ..." In the above case, "return all features that are {family = "wordsent"}, {language = "general"}".

Also, see how setting return_format variable to "list_key" returns a list of the feature keys that match the user-given attributes. Now, we pass those searched keys into extract function.

```
# now, extract the handcrafted linguistic features that you need
extracted_features = LFTK.extract(features = searched_features)

# {'t_word': 8, 't_stopword': 4, 't_punct': 1, 't_uword': 9, 't_sent': 1, 't_char': 36}
print(extracted_features)
```

Q: What if I wanted to extract features from multiple groups?

search_features function only allows users to pass one argument per attribute. This means that you will need to make multiple individual calls. For example, to obtain a list of features from wordsent family and readtimeformula family,

```
searched_features_A = lftk.search_features(family = "wordsent", return_format = "list_key")
searched_features_B = lftk.search_features(family = "readtimeformula", return_format = "list_key")
result = searched_features_A + searched_features_B
```

Then, you can call the usual extraction function,

Contributors 3



brucewlee Bruce W. Lee



strickvl Alex Strick van Linschoten



dangne Dang Nguyen

Languages

• Python 100.0%