Week 03 Handson - Data Preprocessing #02 (Text)

To use a learning model, we need to input numerical data to the model. However, we often get non-numerical data as input, e.g., text data. Thus, to use text as input to the learning model, we need to do pre-processing and convert it to numerical data.

Steps below are typical pre-processing steps for text data.

- 1. Tokenization
- 2. Normalization
- 3. Cleaning
- 4. Lemmatization/stemming

Tokenization

```
In [1]: # please install nltk in your python env.
          # if you use pip, you type: pip install nltk
          import nltk
          from nltk.tokenize import word tokenize
          # we have two raw texts here that we want to pre-process
          text1 = "After watching two hours non stop, \
                    he says that the film is really fantastic #brilliant."
          text2 = "Foods sold there are little bit pricy, \
                   meanwhile the taste is not delicious #notrecommended."
          tokens1 = word tokenize(text1)
          print("tokens1:\n", tokens1)
          tokens2 = word_tokenize(text2)
          print("\n\ntokens2:\n", tokens2)
          ['After', 'watching', 'two', 'hours', 'non', 'stop', ',', 'he', 'says',
'that', 'the', 'film', 'is', 'really', 'fantastic', '#', 'brilliant',
          tokens2:
          ['Foods', 'sold', 'there', 'are', 'little', 'bit', 'pricy', ',', 'meanwh ile', 'the', 'taste', 'is', 'not', 'delicious', '#', 'notrecommended',
          '.']
```

Normalization

In this block of code, we try one of normalization processes: converting to lowercase.

```
In [2]: # convert to lower case
    normalized_words1 = [w.lower() for w in tokens1]
    print("normalized_words1:\n", normalized_words1)

    normalized_words2 = [w.lower() for w in tokens2]
    print("\n\nnormalized_words2:\n", normalized_words2)

    normalized_words1:
        ['after', 'watching', 'two', 'hours', 'non', 'stop', ',', 'he', 'says', 'that', 'the', 'film', 'is', 'really', 'fantastic', '#', 'brilliant', '.']

    normalized_words2:
        ['foods', 'sold', 'there', 'are', 'little', 'bit', 'pricy', ',', 'meanwh ile', 'the', 'taste', 'is', 'not', 'delicious', '#', 'notrecommended', '.']
```

Cleaning 01: remove punctuation

```
In [3]: # remove punctuation from each word
import string
  table = str.maketrans('', '', string.punctuation)
  punc_removed1 = [w.translate(table) for w in normalized_words1]
  print("punc_removed1:\n", punc_removed1)

  punc_removed2 = [w.translate(table) for w in normalized_words2]
  print("\n\npunc_removed2:\n", punc_removed2)

punc_removed1:
  ['after', 'watching', 'two', 'hours', 'non', 'stop', '', 'he', 'says',
  'that', 'the', 'film', 'is', 'really', 'fantastic', '', 'brilliant', '']

  punc_removed2:
  ['foods', 'sold', 'there', 'are', 'little', 'bit', 'pricy', '', 'meanwhile', 'the', 'taste', 'is', 'not', 'delicious', '', 'notrecommended', '']
```

Cleaning 02: remove not alphabetic

```
In [4]: # remove remaining tokens that are not alphabetic
    isalpha_words1 = [word for word in punc_removed1 if word.isalpha()]
    print("isalpha_words1:\n", isalpha_words1)

    isalpha_words2 = [word for word in punc_removed2 if word.isalpha()]
    print("\n\nisalpha_words2:\n", isalpha_words2)

isalpha_words1:
    ['after', 'watching', 'two', 'hours', 'non', 'stop', 'he', 'says', 'that', 'the', 'film', 'is', 'really', 'fantastic', 'brilliant']

isalpha_words2:
    ['foods', 'sold', 'there', 'are', 'little', 'bit', 'pricy', 'meanwhile', 'the', 'taste', 'is', 'not', 'delicious', 'notrecommended']
```

Cleaning 03: remove stop words

Stemming

```
In [6]: from nltk.stem import PorterStemmer
    ps = PorterStemmer()

stemmed_word1 = [ps.stem(w) for w in stopWords_removed1]
    print("stemmed_word1:\n", stemmed_word1)

stemmed_word2 = [ps.stem(w) for w in stopWords_removed2]
    print("\n\nstemmed_word2:\n", stemmed_word2)

stemmed_word1:
    ['watch', 'two', 'hour', 'non', 'stop', 'say', 'film', 'realli', 'fantas t', 'brilliant']

stemmed_word2:
    ['food', 'sold', 'littl', 'bit', 'prici', 'meanwhil', 'tast', 'delici', 'notrecommend']
```

Lemmatization

```
In [7]: from nltk.stem import WordNetLemmatizer
lemmatizer = WordNetLemmatizer()

lemmatized_words1 = [lemmatizer.lemmatize(w) for w in stopWords_removed1]
print("lemmatized_words1:\n", lemmatized_words1)

lemmatized_words2 = [lemmatizer.lemmatize(w) for w in stopWords_removed2]
print("\n\nlemmatized_words2:\n", lemmatized_words2)

lemmatized_words1:
   ['watching', 'two', 'hour', 'non', 'stop', 'say', 'film', 'really', 'fan tastic', 'brilliant']

lemmatized_words2:
   ['food', 'sold', 'little', 'bit', 'pricy', 'meanwhile', 'taste', 'delici ous', 'notrecommended']
```

Example of Converting Preprocessed Text into Numerical Features

```
In [ ]: from sklearn.feature_extraction.text import TfidfVectorizer
        # merge two texts into one list (you may also try to use the stemmed word)
        two_preprocessed_text = [lemmatized_words1, lemmatized_words2]
        # define the tfidf vectorizer
        def dummy(doc):
            return doc
        tfidf = TfidfVectorizer(
            analyzer='word', #'
            tokenizer=dummy,
            preprocessor=dummy,
            token pattern=None)
        # train / Learn from the given data
        model = tfidf.fit(two_preprocessed_text)
        # transform to numerical features using the trained model
        numerical_features = model.transform(two_preprocessed_text).toarray()
        """ --> these numerical features can then be used by the model,
                e.g., for classification to sentiment class: positive and negative
        print("numerical features of text1:\n", numerical features[0],
               "; shape:", numerical_features[0].shape)
        print("\n\nnumerical_features of text2:\n", numerical_features[1],
               "; shape:", numerical_features[1].shape)
```

Question 01 (Q01)

- a. What are tokenization, normalization and cleaning?
- b. What is/are the difference(s) between stemming and lemmatization?

Answer:

[write your answers here, can use Bahasa]

Q02

Please explain what TF-IDF is!

Note: (i) you can insert picture (if you want) in the answer, and then upload all the materials (this ipynb file and the pictures) into one zip file to the course portal, (ii) you can also use mathematical equation here, for exampe: you can write $log_2(P_i)$ by using $\log_{2}(P_{i})$.

Answer:

[write your answer here]

(Bonus) Q03

What are other methods that can be used to convert "preprocessed text" to "numerical features" other than TF-IDF? From what you mention, what are methods that keep the semantic?

Answer:

[write your answer here]